

# Genesis: Data Plane Synthesis in Multi-Tenant Networks

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**Abstract—** Operators in multi-tenant cloud datacenters require support for diverse and complex end-to-end policies, such as, reachability, middlebox traversals, isolation, traffic engineering, and network resource management. We present Genesis, a data-center network management system which allows policies to be specified in a declarative manner without explicitly programming the network data-plane. Genesis tackles the problem of enforcing policies by synthesizing switch forwarding tables. It uses the formal reasoning foundations of constraint solving in combination with fast off-the-shelf SMT solvers. To improve synthesis performance, Genesis incorporates a novel search strategy that uses regular expressions to specify properties that leverage the structure of datacenter networks, and a divide-and-conquer synthesis procedure which exploits the structure of policy relationships. We have prototyped Genesis, and conducted experiments with a variety of workloads on real-world topologies to demonstrate its performance.

## 1. Introduction

Many enterprises are increasingly migrating their on-premise IT infrastructure to cloud datacenters. In such environments, the different enterprises (tenants) share different resources, such as, the compute machines that run their applications and network infrastructure used for communication among these applications. Operators of such multi-tenant datacenters thus have to deal with a multitude of machines communicating with each other (flows) over a network that is composed of many tens to hundreds of routers or switches (devices) [10]. With growing diversity of enterprise applications and the need for security and compliance, these pathways of communication through the datacenter network are subject to increasingly complex network-based policies.

Consider a tenant in such a datacenter. She may desire basic communication among her applications (reachability) along shortest paths based on certain metrics. In addition, she may wish that traffic attempting to reach some of her applications is examined by a set of “middleboxes” (traversal) for auditing and access control. For strong security or Quality-of-Service considerations, a tenant may additionally desire that a subset of her flows does not share any infrastructure with others’ flows (isolation). In parallel, cloud operators must meet key operational policies. For instance, they often need to optimize network performance objectives (traffic engineering), e.g., minimizing the maximum load imposed by all tenants on network links, and deal with resource constraints such as link capacity bounds and switch table sizes. Also, since datacenter networks are highly prone to link/switch failures [11], operators need to gracefully transition the old (pre-failure) data plane to a policy-compliant one (post-failure) in a rapid and/or efficient manner.

Today, configuring network devices to enforce these complex policies in aggregate is manual, ad-hoc, and error-prone. This can lead to misconfigurations and violations of tenant service-level agreements, which can have severe performance and security impacts.

With software-defined networking (SDN), operators can program networks in a more intuitive manner. In SDN, a general-purpose centralized controller machine (control plane) controls end-to-end communication pathways by managing network forwarding rules on a collection of programmable switches (data plane). Using a global view of the current network topology, the controller can program forwarding rules on switches based on application requirements. Unfortunately, existing SDN programming languages (e.g., Frenetic [9] and Pyretic [18]) present too narrow a view: operators would ideally want to specify and realize policies network-wide, whereas these languages focus on programming *individual* switch behaviors. Moreover, for many types of policies, generating a data plane that enforces them is a computationally hard problem, requiring the design of efficient custom heuristics per policy type. Other recent works on network-wide policy enforcement [22, 25] go beyond the single-switch model, but they target specific types of policies and thus are difficult to extend to other commonly desired policy types (e.g., isolation).

In this paper, we seek a general approach that allows a variety of rich policies to be specified as the input, with the output being the corresponding set of switch forwarding rules such that the complexities of correctly realizing the policies in the data plane are hidden from operators. This is an important step toward *intent-based networking* [2], where operators specify *what* they want the network to do instead of worrying about *how* the network must be configured. We argue that data plane synthesis can help realize this vision in the multi-tenant datacenter context.

We present Genesis, a framework for declaratively specifying and enforcing complex policies such as, isolation, middlebox traversals, network optimization objectives, and failure resilience. To tackle the high complexity of enforcing some of these policies (e.g., enforcing isolation is NP-complete), Genesis encodes the problem of enforcing policies as a constraint solving problem and leverages recent advances in fast Satisfiability Modulo Theories (SMT) solvers to efficiently search for a solution to the constraints. The solution is then translated into switch forwarding rules. Genesis uses two intuitive relations that concisely capture the semantics of custom network forwarding behaviors. These help express a variety of both path-based and global policies desired in a datacenter. Interestingly, complex global policies (specifically, policy-compliant failure resilience) can be realized within this framework without requiring additional encoding (of specific failure scenarios) by just cleverly transforming path-based policies. By leveraging the formal guarantees of constraint solving, Genesis eliminates the room for error in the enforcement of complex policies.

Further, we present two novel techniques that leverage domain-specific properties to speed up Genesis’s synthesis. First, Genesis allows the network operator to write restricted forms of regular expressions, called *tactics*, that blacklist paths based on certain patterns that are not desired in a datacenter network (e.g., paths that alternate between topology tiers). These tactics are used to discard several constraints, acting as a search strategy for the

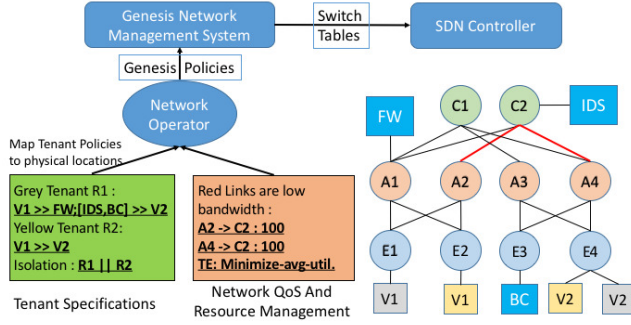


Figure 1: Genesis in a multi-tenant datacenter setting of a network containing several VMs and middleboxes. The network operator translates the tenant specifications and network resource policies to Genesis policies and Genesis synthesizes switch forwarding rules which are installed by the SDN controller.

solver. Tactics can speed up the synthesis procedure by  $1.5 - 400 \times$  (median speedup:  $1.6 \times$ , average speedup:  $22 \times$ ). Second, we develop a *divide-and-conquer* synthesis procedure that opportunistically leverages the dependency relationships among isolation policies to improve synthesis performance. The procedure partitions the input policies into components such that Genesis can synthesize these components separately and faster than the complete problem. Divide-and-conquer synthesis can halve the synthesis time for 40% of synthetic isolation workloads.

**Contributions.** Our contributions are the following.

- An extensible declarative framework for describing complex policies and a modular SMT-based algorithm for enforcing policies like isolation, waypoints (§4), traffic engineering (§5), and failure resiliency (§5.3.1);
- A modified synthesis algorithm based on tactics, which leverages datacenter network structure to blacklist undesirable path patterns (§6);
- A divide-and-conquer procedure for speeding up synthesis by leveraging the structure of policy interactions (§7);
- An implementation of Genesis and an extensive evaluation on different policy workloads, topologies and multi-tenancy settings (§8).

A long version of this paper containing all the proofs has been submitted as supplementary material.

## 2. Preliminaries and Policies Supported

We describe the type of policies desired in multi-tenant data centers that Genesis supports. We use Figure 1 as a running example. In our setting, an “operator” manages a multi-tenant datacenter, which could be either a public datacenter (E.g., Azure), or a private datacenter. The operator specifies “operator policies” which reflect important global objectives pertaining to how she wishes to manage her overall infrastructure. A tenant is an entity (e.g., an enterprise or a department thereof) that has offloaded its IT infrastructure to the datacenter. Each tenant controls a number of host machines in the datacenter running some of its applications, and specifies path-based policies (as opposed to operator’s global policies). Tenant policies define whether paths can exist among its hosts, and if so, what additional properties the paths must satisfy for security, performance or access control reasons. Figure 1 shows several tenants who differ in the nature of policies they wish to realize; we also show the operator policies.

The policies supported by Genesis are described below. Notice that these reflect and, in some cases, extend policies that today’s enterprises and datacenter operators realize in their networks [10].

- **Tenant Policy: Reachability.** This enables network communication between specific pairs of tenant’s virtual instances (VMs), applications, or hosts. In our example, one tenant has defined a reachability policy for its two VMs (R2):  $V1 \gg V2$ , which is translated after VM placement (which is handled transparently by the operator) as  $E2 \gg E4$ . A pair of VMs, applications or hosts that are allowed to communicate by means of a reachability policy defines a flow or a “packet class”; we use these terms interchangeably. Any communication that is not defined by a reachability policy is implicitly blocked (i.e., all communication is “default off”).
- **Tenant Policy: Middlebox Traversals.** A tenant may wish that the flow between two of her end hosts, or from another tenant, must traverse specific middleboxes, which we also refer to as “waypoints” in this paper. Middleboxes are custom processing appliances often used for security, access control, or performance reasons (e.g., firewalls, intrusion prevention systems, monitoring/accounting gateways, proxies, and load balancers). Specifically, for particular flows of interest, a tenant can provide a sequence of unordered sets of middleboxes to traverse [21]. The flows must traverse these sets in order, while in a set, all middleboxes must be traversed and the order is irrelevant.<sup>1</sup> For example, one of the tenants defines a traversal policy (R1):  $V1 \gg FW; [IDS, BC] \gg V2$  which specifies that traffic must first pass through the firewall (FW), and then through the Intrusion detection system (IDS) and the byte counter (BC) in any order.
- **Tenant Policy: Isolation.** Tenants may require various Quality-of-Service (QoS) or security guarantees that stipulate varying degrees of isolation for their traffic. In the extreme, a tenant could require that her flows are not affected in any manner by any other tenant by strictly isolating the path of the tenant’s flows from others’ flows. In Figure 1, we have two tenants whose traffic will be isolated from one another ( $R1 || R2$ ), i.e., the network paths used by the tenants will not share any links in the topology. A tenant could also specify isolation for a subset of her (performance-sensitive) flows from other flows of the same tenant or those belonging to other tenants; the rest of the tenant’s flows may require no guarantees.
- **Operator Policy: Managing Capacity Constraints.** While support for the above policies can be used to satisfy tenant requirements, network operators often wish to carefully managing constrained resources. Common examples that Genesis supports include enforcing strict constraints on aggregate number of flows traversing a switch (due to all tenants) so as to adhere to switch memory constraints, and ensuring that the total load on certain links is within predefined thresholds (we assume here that each flow has a predefined load that it imposes on the path it uses). In our example in Figure 1,  $A2-C2$  and  $A4-C2$  links are of low bandwidth, and the operator wants to ensure that total load on these links does not exceed 100 ( $A2 \rightarrow C2 : 100, A4 \rightarrow C2 : 100$ ).
- **Operator Policy: Traffic Engineering.** As an alternative to managing strict (link) capacity constraints, operators may also want to balance load on their network infrastructure. This is often done by optimizing a network-wide objective such as total or maximum utilization of network links due to traffic induced by all tenants.
- **Operator Policy: Handling failure gracefully.** Modern networks experience link and switch failures frequently [1]. When a failure occurs, we must reconfigure the forwarding rules so

<sup>1</sup> The unordered set abstraction leverages the fact that middleboxes without dependencies in their traffic processing behavior can be placed in any order relative to each other [21].

Type	Policy	GPL Syntax	Description
Tenant	Reachability	$predicate : src \gg dst$	Forwarding Rules for path from switch $src$ to switch $dst$ for packets matching $predicate$
	Reachability with Ordered Waypoints	$predicate : src \gg W_1; \dots; W_n \gg dst$	Forwarding Rules for path from switch $src$ to switch $dst$ for packets matching $predicate$ such that the path traverses $w \in W_1$ in any order, then $w \in W_2$ in any order after all waypoints in $W_1$ are traversed and so on.
	Traffic Isolation	$R1 \parallel R2$	Paths of two reachability policies $R1$ and $R2$ do not share a link in the same direction
	Link Isolation	$R1 \lt \gt R2$	Paths of two reachability policies $R1$ and $R2$ do not share a link in any direction (edge-disjoint)
Operator	Link Capacity	$sw_1 \rightarrow sw_2 : capacity$	The weights of flows traversing the link $sw_1 \rightarrow sw_2$ do not exceed $capacity$
	Switch Table Size	$sw : size$	The number of flows traversing through $sw$ do not exceed $size$ as each flow would require a forwarding rule at $sw$
	Traffic Engineering	$minimize - tot - te, minimize - max - te$	TE objectives: minimize total/max link utilization

Table 1: Genesis Policy Support with Genesis Policy Language (GPL) syntax

that the policies are satisfied. Naively recomputing forwarding rules incurs an unduly large overhead because old forwarding rules have to be torn down at all switches, and new rules must be installed. Switch rules deletions/insertions take a non-trivial amount of time [? ], potentially leading to disruptions. It is therefore desirable to have graceful approaches that either minimize the potential for disruption by minimizing the number of forwarding rules or switches modified in transitioning from an old data-plane, or eliminate the possibility of disruption altogether by precomputing backup policy-compliant paths for a fixed number of failures (at the cost of storing extra rules at switches).

Realizing these policies is challenging today. In particular, state-of-the-art SDN frameworks, e.g., Pyretic [ ] and Frenetic [ ], are insufficient to program networks to realize them. This is because the above policies are global and cannot be enforced (at least not in an intuitive manner) by programming individual behavior of switches. While some existing SDN-based network management systems [15, 22, 25] overcome these limitations by taking a network-wide view, they are tailored to support specific policies such as middlebox placement or link capacity constraints. As such they cannot enforce several of the policies above.

### 3. Data Plane Synthesis

Our contribution is Genesis, a new general network management system that supports the above policies, and can be extended to support others. The architecture of Genesis is shown in Figure 1. Genesis performs *synthesis* of switch forwarding rules to enforce policies. The policies are specified using Genesis Policy Language, or GPL, as shown in Table 1.

Unlike previous efforts in the network synthesis space [23, 25], Genesis is not tailored to specific formalisms such as regular expressions; this aspect makes it *modular* and *easy to extend*. To draw an analogy with SMT solvers, Genesis can be seen as a constraint solver that allows the addition of different types of policies (respectively, theories in SMT) and the design of optimizations based on properties desired by network operators.

Our work is motivated by recent advances in program synthesis, i.e., the task of discovering an executable program from user intent expressed in the form of some constraints. There are three key dimensions to a synthesis problem: the type of constraints that it accepts as expression of user intent, the space of programs over which it searches, and the search technique it employs. Genesis leverages synthesis as follows: given a set of policies which describe tenant and operator intent, the search space is the space of all data planes (i.e., the set of forwarding rules) and the search technique involved is SAT/SMT solving.

Genesis’s approach has the following salient features:

(1) Enforcement of the different policies can be translated to the following problem: Given a set of node pairs (derived from the reachability policies) in the graph (topology), find paths in the graph for each of the node pairs satisfying certain properties (derived from the rest of the policies). Thus, the different policies can be enforced by a correct set of forwarding rules at the switches. No extra functionality is required from the controller; its only role is to install the forwarding rules on switches.

(2) Correct enforcement is challenging due to different goals for each of the policies — ensuring isolation between paths may lead to overshooting link utilizations and vice-versa — and is a common cause of incorrect configurations in networks. Our approach removes the need for a verification step in which the operator has to “check” whether the forwarding rules satisfy the desired policies. By using a formal reasoning technique, we are able to consider the space of all data planes and find a solution which is *correct by construction*, eliminating room for operator errors.

(3) Automatically enforcing policies is a task with *high theoretical complexity*. For example, enforcing isolation policies is as hard as solving graph-coloring, an NP-complete problem. Specialized techniques can be used to find the forwarding rules when handling a particular class of policy, but devising good search techniques becomes challenging when multiple types of policies are combined, — e.g., isolation, waypoints, and traffic engineering. Thanks to the many engineering efforts, SMT solvers abstract away most of this complexity and allow us to unify search objectives for every policy into a generalized search technique. Crucially, Genesis can be extended with ease to support new policies without requiring changes to the underlying search techniques.

In the next section, we describe the Genesis synthesis algorithm for tenant policies. We then describe how to accommodate operator policies pertaining to capacity constraints, traffic engineering and failure resiliency (§5). Finally, we describe two novel techniques aimed at speeding up Genesis’s synthesis: tactics (§6) and divide-and-conquer synthesis (§7).

### 4. Synthesis of Tenant Policies

The problem statement here is as follows: Given the network topology graph and the set of tenant policies written in GPL, generate paths in the network for every source-destination pair (derived from reachability policies) satisfying all policies. To achieve this, Genesis creates constraints that encodes the forwarding and reachability rules pertaining to the paths such that they satisfy the input policies. The synthesized solution of paths obtained from the constraints are then translated to switch forwarding rules.

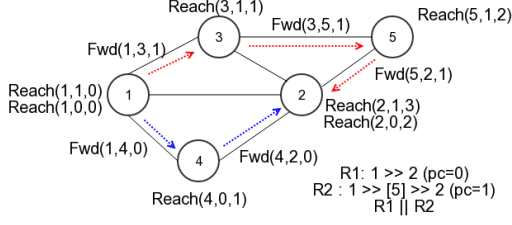


Figure 2: Values of the  $Fwd$  and  $Reach$  relations of the network forwarding model for the policies specified in the figure. The blue and red arrows indicate the paths of packet classes 0 and 1 respectively according to the model.

#### 4.1 Network Forwarding Model

We start by describing the basic forwarding model we use in Genesis. We define the physical switch topology as an undirected graph  $T = (S, L)$ , where  $S$  is the set of switches and  $L$  is the set of links. We use the neighbour function  $N(s) = \{s' \mid (s, s') \in L\}$  to denote the set of neighbour switches of  $s$ . We assume a set of packet classes  $PC : [0, \lambda]$  and map each reachability policy to a unique integer in  $PC$ . In the rest of the paper, we often use the term packet class to identify the corresponding reachability policy. Other policies are not mapped to packet classes as they do not produce a path, but specify restrictions on paths of packet classes. We use  $R$  to denote the set of reachability policies; each policy  $r \in R$  is a pair ( $predicate:src \gg W_1; W_2; \dots W_n \gg dst, pc$ ) where:

- $predicate$  is the packet header identifier pertaining to  $r$ ;
- $src, dst \in S$  are the source and destination switches;
- $W_1, W_2, \dots, W_n \subseteq S$  are the (potentially empty) ordered sets of waypoints;
- $pc \in PC$  is the packet class and is a unique integer used to identify the variables associated to  $r$

We fix a constant  $\mu$  and assume all paths to have length at most  $\mu$ .  $K = [0, \mu]$  is the set of all permissible path lengths. The network forwarding model abstracts the actual forwarding rules at each node and encodes the reachability of each packet class.

We use the relation  $Fwd \subseteq S \times S \times PC$  to capture the network forwarding behavior,—i.e.  $(sw_1, sw_2, pc) \in Fwd$  if  $sw_1$  forwards packets of class  $pc$  to switch  $sw_2$ . We use the relation  $Reach \subseteq S \times PC \times K$  to capture the path reachability,—i.e.  $(sw, pc, k) \in Reach$  if the switch  $sw$  is reachable in the path from the source switch of packet class  $pc$  in exactly  $k$  steps. For brevity, we write  $Fwd(sw_1, sw_2, pc)$  for  $(sw_1, sw_2, pc) \in Fwd$  and similarly for the  $Reach$  relation. Since  $Fwd$  depends on the topology, for all  $sw_1, sw_2$  that are not connected by a link, we have that  $\forall pc, (sw_1, sw_2, pc) \notin Fwd$ .

Given a set of policies, Genesis generates a set of constraints denoted by  $\Psi$  over the  $Fwd$  and  $Reach$  relations.  $(Fwd, Reach) \models \Psi$  denotes that  $Fwd$  and  $Reach$  is a model of  $\Psi$ .

**Definition 1.** Given two concrete relations  $Fwd$  and  $Reach$ , the set of induced paths  $\Pi = paths(Fwd, Reach)$  is defined as follows: given a class  $pc$ ,  $(sw_0 \dots sw_k, pc) \in \Pi$  iff:

1.  $\forall i \in [0, k]. (sw_i, pc, i) \in Reach$
2.  $\forall i \in [0, k-1]. (sw_i, sw_{i+1}, pc) \in Fwd$

Figure 2 illustrates these definitions.

**Definition 2.** Given the set of constraints  $\Psi$  corresponding to the input policies, a set of paths  $\Pi$  is a solution to  $\Psi$ ,  $\Pi \models \Psi$ , if there exists  $Fwd, Reach$  such that  $(Fwd, Reach) \models \Psi$  and  $\Pi = paths(Fwd, Reach)$ .

In practice, we model the forwarding and reachability relations using propositions and reduce enforcement of tenant policies like

reachability, waypoints and isolation to a Boolean Satisfiability Problem (SAT) problem<sup>2</sup>. Using these relations, operators can write custom policies in a concise and intuitive manner.

#### 4.2 Reachability

We first discuss the constraints added to  $\Psi$  for reachability policies without waypoints. For a reachability policy  $s \gg d$  and packet class  $pc$ , the added constraints must ensure that the solution model represents a path from source to destination. The base constraint states that  $(s, pc, 0) \in Reach$  meaning that  $s$  can be reached in 0 steps. The following constraint states that there must be a forwarding rule from  $s$  to one of the neighbors of  $s$ <sup>3</sup>.

$$\exists n \in N(s). Fwd(s, n, pc) \wedge Reach(n, pc, 1). \quad (1)$$

Next, we add the following constraints that state that  $d$  can be reached in some number of steps and, since  $d$  is the last switch in the path, there are no forwarding rules from it.

$$\exists k. Reach(d, pc, k) \wedge \forall n \in N(d). \neg Fwd(d, n, pc). \quad (2)$$

Finally, we add implication constraints that propagate reachability backward from destination to source. If a node  $n_1$  is reachable in  $k$  steps, there must be a node  $n_2$  reachable in  $k-1$  steps and a forwarding rule  $n_2 \rightarrow n_1$ .

$$\forall n_1, k. Reach(n_1, pc, k) \implies \exists n_2. n_2 \in N(n_1) \wedge Reach(n_2, pc, k-1) \wedge Fwd(n_2, n_1, pc). \quad (3)$$

When combined together, these constraints are sufficient to ensure the existence of a path from  $s$  to  $d$  for packet class  $pc$ . However, since there is no restriction on the number of  $Fwd$  values that can be true at a switch, we can get multiple forwarding rules at switches, and also multiple paths to the destination. These can also create forwarding loops. Concretely, this is not a problem: as long as there is at least one path from  $s$  to  $d$  we can recover it from the solution of the constraints. Moreover, this representation is quite efficient, as forcing a single path would require adding further constraints (§4.3) and increase the synthesis time.

To extract a concrete  $s$ -to- $d$  path we perform a breadth-first search on the reachability graph induced by the solution to the constraints. A directed edge  $n_1 \rightarrow n_2$  appears in the reachability-graph if there is a forwarding rule indicated by the relation  $(n_1, n_2, pc) \in Fwd$ . We extract the rules relevant to the shortest path from source to destination from the model, and the additional rules obtained in the solution (extra paths, forwarding loops) are ignored.

#### 4.3 Waypoint Traversal

For a reachability policy with a sequence of waypoint sets  $s \gg W_1; \dots; W_n \gg d$  and packet class  $pc$ , we add all the constraints specified in §4.2 to ensure the existence of a path from  $s$  to  $d$ . We then add constraints so that all waypoints  $w$  are traversed.

$$\forall w \in W_1, \dots, W_n. \exists k. Reach(w, pc, k). \quad (4)$$

For each set  $W_i$  for  $i > 1$ , we add constraints to ensure that all waypoints in  $W_i$  are reached after all waypoints in  $W_{i-1}$ :

$$\forall w_i \in W_i, \forall k_i. Reach(w_i, pc, k_i) \implies \forall w_{i-1} \in W_{i-1}. \exists k_{i-1}. k_{i-1} < k_i \wedge Reach(w_{i-1}, pc, k_{i-1}). \quad (5)$$

<sup>2</sup> An earlier iteration of our model used *uninterpreted functions* and modeled reachability using recursive constraints, which was slower with a greater number of constraints.

<sup>3</sup> We unroll the existential quantifier  $\exists n \in N(s)$  using disjunction of clauses  $\bigvee_{n \in N(s)}$  and the universal quantifier  $\forall n \in N(dst)$  using conjunction of clauses  $\bigwedge_{n \in N(dst)}$  and stay in propositional logic.

Previously, we imposed no restriction on the number of paths from  $s$  to  $d$ . In the case of waypoints, this can result in a solution with multiple paths, with each individual path traversing some of the waypoints, which is not the correct enforcement for a waypoint policy. Thus, we need to ensure the solver returns a single path traversing all the waypoints. To achieve this, we limit the number of forwarding rules for  $pc$  at a switch to 0 or 1. We define the forwarding set as:

$$FwdSet(sw, pc) = \{k \mid Fwd(sw, k, pc)\}. \quad (6)$$

We then add constraints stating that the size of the forwarding set must not exceed 1:

$$\forall sw, pc. |FwdSet(sw, pc)| \leq 1. \quad (7)$$

Here  $|A|$  denotes the size of set  $A$ . The above constraints are expressed in SAT as follows:

$$\bigvee_{k_1 \in N(sw)} Fwd(sw, k_1, pc) \wedge \left( \bigwedge_{k_2 \in N(sw), k_2 \neq k_1} \neg Fwd(sw, k_2, pc) \right) \quad (8)$$

Since, there cannot exist multiple rules at a switch, the model will contain a single path from source to destination for  $pc$  traversing the waypoints in the right order.

#### 4.4 Isolation

A traffic isolation policy  $pc_1 || pc_2$  states that the paths for  $pc_1$  and  $pc_2$  do not share any link in the same direction. We enforce this policy by adding to  $\Psi$ , constraints stating that at every switch,  $pc_1$  and  $pc_2$  must not forward to the same switch:

$$\forall n_1. \neg(\exists n_2. Fwd(n_1, n_2, pc_1) \wedge Fwd(n_1, n_2, pc_2)). \quad (9)$$

For a link isolation policy  $pc_1 <> pc_2$  which prevents sharing a link in both directions, we add the constraints:

$$\forall n_1. \neg(\exists n_2. Fwd(n_1, n_2, pc_1) \wedge (Fwd(n_1, n_2, pc_2) \vee Fwd(n_2, n_1, pc_2))). \quad (10)$$

With single paths for  $pc_1$  and  $pc_2$  (when combined with Equation (7)), the above constraints ensure those paths are isolated. Interestingly, for a reachability policy without waypoints, the constraints in Equation (7) are not required to enforce isolation. Even though the solver could produce multiple forwarding rules which induce multiple paths, the constraints in Equation (9) or Equation (10) guarantee isolation as the solver would discard the rules conflicting with another packet class.

## 5. Operator Policies

We now describe how to extend Genesis's synthesis to support various operator policies. We describe Genesis can support hard capacity constraints and optimization objectives pertaining to traffic engineering using linear rational arithmetic (LRA) and linear optimization objectives in SMT. We conclude by describing how Genesis can be extended to allow operators to handle datacenter network failures in a graceful policy-compliant manner.

### 5.1 Link and Switch Table Capacity Constraints

For a link capacity policy on the link  $sw_1 \rightarrow sw_2 : \omega$ , Genesis must ensure that the sum of traffic rates of packet classes using link  $sw_1 \rightarrow sw_2$  does not exceed  $\omega$ . As input, we have the traffic rates  $\sigma(pc)$  of each of the packet classes. The constraints added to  $\Psi$  are:

$$\sum_{\forall pc} ite(Fwd(sw_1, sw_2, pc), \sigma(pc), 0) \leq \omega. \quad (11)$$

If a class  $pc$  uses link  $sw_1 \rightarrow sw_2$ , then  $(sw_1, sw_2, pc) \in Fwd$  and  $\sigma(pc)$  is added in the utilization of the link.

A switch table policy  $sw : \gamma$  specifies that the number of forwarding rules on  $sw$  must not exceed  $\gamma$ . Similar to the link capacity policy, the constraints ensure the count of all packet classes which traverse  $sw$  (each will require a forwarding rule) is  $\leq \gamma$ :

$$\sum_{\forall pc} ite(\exists k. Reach(sw, pc, k), 1, 0) \leq \gamma. \quad (12)$$

### 5.2 Traffic Engineering

While the above capacity policies can be used to perform a strict form of traffic engineering (TE) in terms of adhering to link bandwidths, it is often more useful to balance traffic across links because a link failure will affect fewer flows when the flows are spread evenly across the network. To this end, network operators often impose traffic engineering objectives such as minimizing the total link utilization or the maximum link utilization.

**Min-tot TE** To perform traffic engineering, link capacities of the network  $C(sw_1, sw_2)$  and traffic rates of the packet classes  $\sigma(pc)$  are specified as input to Genesis (we assume a single path for a packet class). The utilization of a link  $U(sw_1, sw_2)$  is defined as the ratio of total traffic flowing through the link to the link capacity, and encoded using the theory of linear rational arithmetic as:

$$U(sw_1, sw_2) = \frac{\sum_{\forall pc} ite(Fwd(sw_1, sw_2, pc), \sigma(pc), 0)}{C(sw_1, sw_2)} \quad (13)$$

The following objective minimizes the total link utilization:

$$\text{minimize} \sum_{\forall sw_1, sw_2} U(sw_1, sw_2) \quad (14)$$

**Min-max TE** To encode the TE objective of minimizing the maximum link utilization, we define a variable  $maxU$  which represents the maximum link utilization. The constraints added to ensure that  $maxU$  is greater than or equal to all individual link utilizations:

$$\forall sw_1, sw_2. maxU \geq U(sw_1, sw_2) \quad (15)$$

We then impose the following objective:

$$\text{minimize} maxU \quad (16)$$

Using an encoding similar to the one presented in this section, Genesis can be used for other quantitative objectives like minimizing total latency and load balancing traffic across middleboxes.

### 5.3 Handling Failures Gracefully

Another network management consideration for operators is the occurrence of failures (switches, links etc.), which are all too frequent in datacenter networks [11]. Failures require recomputation of paths compliant to the input policies for the modified topology. A naive approach is to use Genesis to resynthesize the modified instance; However, the new solution may be drastically different from the original data plane, incurring a large overhead of removing old rules and installing new ones [13, 14].

In what follows, we describe two techniques to handle failures more gracefully. The first technique is data-plane resiliency (§5.3.1), which synthesizes and pre-installs resilient data planes, which even in the event of a bounded number of link failures, continue to satisfy input policies. This technique eliminates the need to resynthesize the forwarding rules for every network failure event, but it requires extra backup rules on switches, and it also cannot capture global operator policies.

Thus, we propose a second mechanism called *minimal repair* (§5.3.2), which can transition from the disrupted data plane to a new policy-compliant one with minimal overhead by minimizing the number of switches whose rule tables are modified. Repair does not incur the extra rule cost of the first approach and can capture all



Genesis policies. It is also useful for accommodating incremental policy changes, which occur frequently in cloud datacenters [10]. The main drawback is that it still requires removal/installation of rules when a failure occurs, which can end up being expensive depending on the number of switches involved.

### 5.3.1 Dataplane Resiliency

In this section, we describe the transformation of input policies to provide dataplane *t-resilience* [26], i.e., in the event of upto  $t$  arbitrary link failures, the synthesized data plane still has a path for each packet class satisfying all policies which is achieved by synthesizing backup paths that satisfy input policies. We only consider reachability, waypoint, and isolation policies in the input. Global policies like capacity policies and traffic engineering pose a difficulty in synthesis. For example, consider a packet class  $pc$  with a traffic rate of  $\sigma(pc)$ . By considering the backup paths with the same traffic characteristics for synthesis, the total traffic accounted for  $pc$  would be  $c \times \sigma(pc)$  (for some constant  $c$ ), leading to under-provisioning of resources. Our current resilience transformation has no provisions to avoid or minimize the under-provisioning of resources which affect capacity policies and TE objectives.

Given the physical topology  $T = (S, L)$ , we define a link-failure scenario  $\theta$  as the set of failed links such that  $\theta \subseteq L$ . We define  $\Theta(t)$  as the set of all failure scenarios where no more than  $t$  arbitrary links fail,—i.e.  $\Theta(t) = \{\theta \mid |\theta| \leq t\}$ . Given a packet class  $pc$ , we construct the induced data plane graph  $\xi = (S, L_{pc})$  from the links of the paths returned by the synthesis algorithm for class  $pc$ . For a failure scenario  $\theta$ , the active data plane  $\xi_\theta = (S, L_{pc} \setminus \theta)$  represents all the links used by  $\xi$  which are unaffected by the failure scenario. A data plane  $\xi$  is *resilient* to  $\theta$  if it contains a path from the source to destination for the packet class in the active data plane  $\xi_\theta$ .

**Definition 3 (Resilience).** A data plane  $\xi = (S, L_{pc})$  for class  $pc$  is *t-resilient* if  $\xi$  is resilient to all  $\theta \in \Theta(t)$ .

While resilience deals with existence of paths during failure scenarios, we extend the notion to include policy compliance.

**Definition 4 (Policy-compliance).** A *t-resilient* data plane  $\xi = (S, L_{pc})$  for class  $pc$  is *policy-compliant* if under any failure scenario  $\theta \in \Theta(t)$ , any path for  $pc$  in  $\xi_\theta = (S, L_{pc} \setminus \theta)$  satisfies the input policies.

#### Algorithm 1 Resilience Transformation

```

1: [Input]  $PC$ : Packet classes (Reachability/Waypoint policies)
2: [Input]  $I$ : Isolation policies (Traffic and Link types)
3: [Input]  $t$ : Maximum number of arbitrary link failures
4: [Output]  $PC^R, I^R$ : Transformed set of policies such that the synthesized data plane is t-resilient and policy-compliant

5:  $PC^R, I^R \leftarrow \emptyset$ 
6: for  $pc : \{src_{pc}, dst_{pc}, W_{pc}\} \in PC$  do
7:   // Create  $t + 1$  edge-disjoint paths of  $pc$ 
8:    $\hat{pc} = \{rc_1, rc_2, \dots, rc_{t+1}\}$  s.t.  $\forall m. rc_m : \{src_{pc}, dst_{pc}, W_{pc}\}$ 
9:    $PC^R = PC^R \cup \hat{pc}$ 
10:   $I_{pc} = \{rc_m \prec rc_n \mid \forall m, n \leq t + 1 \wedge m < n\}$ 
11:   $I^R = I^R \cup I_{pc}$ 
12: end for
13: for  $i : pc_m < op > pc_n \in I$  do
14:   $\hat{i} = \{rc_1 \prec op > rc_2 \mid \forall rc_1 \in \hat{pc}_m, \forall rc_2 \in \hat{pc}_n\}$ 
15:   $I^R = I^R \cup \hat{i}$ 
16: end for
17: return  $PC^R, I^R$ 

```

Algorithm 1 shows how Genesis can be used to provide *t-resilience*. The idea is to modify the input policies such that multiple disjoint paths satisfying the original policies are synthesized

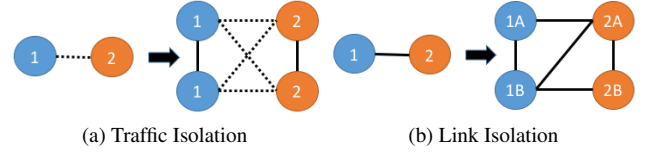


Figure 3: (a) Resilience Transformation for  $pc_1 \parallel pc_2$  for providing  $1 - resilience$ . The dotted lines represent traffic isolation policies, while the solid lines represent link isolation. (b) Example of a sufficient transformation for  $1 - resilience$  in the case of a link-isolation policy.

for each packet class. For *t-resilience*, a packet class  $pc$  needs at least  $t + 1$  edge-disjoint paths from source to destination. We ensure this property holds by creating  $t + 1$  new packet classes ( $\hat{pc}$  in line 8) and use *link-isolation policies* amongst all pairs in  $\hat{pc}$  (line 10) to create  $t + 1$  edge-disjoint paths for  $pc$ . The synthesized data plane  $\hat{\xi} = (S, L_{pc})$  for class  $pc$  is constructed from the paths in the resilient packet class set  $\hat{pc} = \{rc_1, \dots, rc_{t+1}\}$ , i.e.,  $L_{pc} = \bigcup_{rc \in \hat{pc}} L_{rc}$ . Each path of  $\hat{pc}$  satisfies the reachability policy, and any arbitrary  $t$  link failure scenario cannot affect all  $t + 1$  paths.

However, the resilient paths need to satisfy the input isolation policies with other packet classes (which themselves have  $t + 1$  paths for resilience). Thus, for a given policy  $pc_1 \parallel pc_2$ , we add isolation policies to every pair of classes of  $\hat{pc}_1$  and  $\hat{pc}_2$  (line 14). This ensures that any path chosen in the data planes of  $pc_1$  and  $pc_2$  will be isolated from one another, thus providing policy-compliance under any arbitrary  $t - link$  failure scenario. Figure 3(a) demonstrates an example transformation for providing  $1 - resilience$ .

We now that Algorithm 1 is sound.

**Theorem 5.1 (Soundness).** Given input policies  $(PC, I)$ , the data plane  $\hat{\xi}_{pc}$  for every packet class  $pc \in PC$  synthesized from transformed policies  $(PC^R, I^R)$  is *t-resilient* and *policy-compliant*.

If there are no isolation policies in the input, the resilience transformation in lines 8-11 of Algorithm 1 is complete.

**Theorem 5.2 (Completeness).** Given input policies  $(PC, I)$  such that  $I = \emptyset$ , the synthesized data plane  $\xi$  for a packet class  $pc$  is *t-resilient* if and only if it contains  $t + 1$  edge-disjoint paths from source to destination for  $pc$ .

When the original policies contain link-isolation policies, the policies from Algorithm 1 may return *unsat* even when a resilient data plane exists. Specifically, line 14 can add additional policies than is required for resilience. Figure 3(b) shows a transformation required for  $1 - resilience$  with a smaller number of link-isolation policies among different classes of  $pc_1$  and  $pc_2$  than one obtained from Algorithm 1. Consider a failure scenario which disables path of  $pc_{1A}$ . By virtue of the link-isolation policies,  $pc_{1B}$  and  $pc_{2A}$  will be unaffected and can be used as paths for  $pc_1$  and  $pc_2$  respectively, and  $pc_1 \prec pc_2$  holds. Now suppose  $pc_{1B}$  is affected. Similarly,  $pc_{1A}$  and  $pc_{2A}$  can be used as the paths for the original packet classes. The same scenarios hold symmetrically for  $pc_2$ , and thus the resilience transformation can be achieved without adding link-isolation policies amongst all the packet classes.

### 5.3.2 Minimal Repair

Dataplane resiliency imposes high rule storage overhead on switches, and cannot accommodate global policies like link capacity bounds. As an alternative to it, we extend Genesis's synthesis algorithm to perform minimal network repair using MaxSMT.

Formally, the MaxSMT problem is as follows: given a set of formulas  $\Psi_0, \Psi_1, \dots, \Psi_n$  with associated weights  $w_1, \dots, w_n$ , find a

subset  $M \subseteq \{1, \dots, n\}$  s.t. 1)  $\Psi_0 \wedge \bigwedge_{i \in M} \Psi_i$  is satisfiable, and 2) The *award*  $\sum_{i \in M} w_i$  is maximized. The constraints  $\Psi_1, \dots, \Psi_n$  denote *soft* constraints, and the associated weights  $w_i$  encode the award for including  $\Psi_i$  in the satisfying assignment.

We reduce the network repair problem to a MaxSMT problem and use soft constraints to minimize the number of switches on which rules need to be updated. Note that the disadvantage w.r.t dataplane resiliency is that switches still require rule updates, which may take time depending on the number of switches involved.

Let the policy constraints generated by Genesis for the new network state be  $\Psi_0$ , and let  $\overline{Fwd}$  be the present data plane that does not satisfy  $\Psi_0$ . The objective is to find new  $Fwd$  which satisfies  $\Psi_0$  while maximizing the number of *preserved switches* (switches whose rules are unchanged). If the rules on switch  $sw_i$  are preserved, then  $Fwd$  and  $\overline{Fwd}$  have the same forwarding rules for all packet classes which traverse through  $sw_i$ . The following soft constraints capture this idea:

$$\Psi_{sw_i} = \bigvee_{\substack{\forall sw_j, pc \\ (sw_i, sw_j, pc) \in \overline{Fwd}}} Fwd(sw_i, sw_j, pc) \quad w_{sw_i} = 1 \quad (17)$$

The solution to this MaxSMT problem is a data plane that minimizes the number of switches whose rules have to be changed. Alternate repair objectives like minimizing the number of changed forwarding rules can be expressed similarly. Interestingly, the Genesis's network repair mechanism can also be used to transform an existing non-compliant data plane to a policy-compliant one.

## 6. Tactics

Synthesis for a reachability policy translates to choosing a path from the solution space of all paths from the source to destination such that the chosen path satisfies all policies, e.g., waypoint traversal and isolation. Datacenter topologies, e.g., fat-trees [4], have numerous paths between edge switches to provide full bisection bandwidth. Thus, the solution space of paths for a pair of endpoints is large. For example, consider the fat-tree topology in Figure 4. The number of paths under length 10 between two edge switches in the same pod is 242 and two edge switches in different pods is 272. If we consider the synthesis of  $n$  packet classes, the problem roughly translates to finding a solution in the space of size  $n \times 242$ . Op-

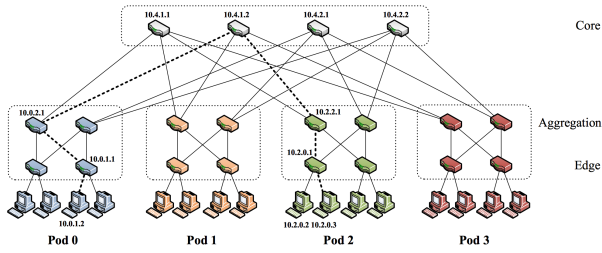


Figure 4: Fat Tree Topology

erators can leverage the network structure of topologies to reduce the solution space by specifying undesirable path patterns. For example, one pattern is that a path between two edge switches in a fat-tree must not traverse another edge switch. This pattern will not drastically reduce the solution space of paths due to the dense interconnect between aggregate and core switches, ensuring that synthesis most likely will not fail.

We introduce *tactics* (the name is inspired from the usage in SMT solvers, not proof assistants) on labels; abstractions that allow

a network operator to impose restrictions on the paths. We use the notion of mapping the set of switches to labels to have a coarse-grained way for specifying path patterns. This helps to create search strategies which can be used for a majority of the packet classes instead of individual switch-level patterns which lack generality.

Let  $Lb$  be the set of labels and  $S$  be the set of switches in the topology. Let  $\phi : S \rightarrow Lb$  be the labeling function that maps each switch to a label in  $Lb$ . One example for  $\phi$  in the fat-tree topology in Figure 4 can be that we map all switches in the same level (core, aggregate or edge) to the same label, therefore leveraging the hierarchical structure of the topology. A path  $p$  is a word over the alphabet  $S$ . We define the path-labeling function  $\Phi : S^* \rightarrow Lb^*$ , which maps each switch in the path to its corresponding label. For example, given the path  $p = e1\ a2\ c3\ a4\ e2$ , the path-labeling function produces  $\Phi(p) = eacae$  – map each switch to its corresponding label. Here,  $e$ ,  $a$ , and  $c$  stand for edge, aggregate, and core respectively.

### 6.1 Synthesis with Tactics

Tactics are simple regular expressions over the set of labels and are used to blacklist certain path patterns. Regular expressions have been previously used in tools like NetGen [23] to specify the paths for a packet class. While supporting full regular expressions is possible, it causes a blow-up in the solving time as further constraints need to be added to the solver to ensure that the path satisfies the regular expression. Instead, using the notion of switch labels, we use regular expressions to provide high-level path properties as a *search strategy* rather than a policy requirement. Rather than specifying how the path must look like, we use regular expressions on switch labels to specify *blacklists* i.e., what the path must *not* look like. A tactic, for example, can blacklist paths from an edge switch to an edge switch that go through another edge switch.

#### 6.1.1 Restricted tactic syntax

For specifying tactics, we identified a restricted set of regular expressions<sup>4</sup> that not only do not require extra constraints to be added, but actually allow us to reduce them. This set is defined using the following grammar:

$$\begin{aligned} R &::= \neg(l_{src}.^i C.^i l_{dst}) \\ C &::= \varepsilon \mid l_1 \mid l_1 l_2 \end{aligned}$$

where  $l_i \in Lb$  and  $l_{src}, l_{dst}$  are used to specify the labels of the source switch and destination switch, respectively, of the packet classes to which we apply the tactic. Since our goal is to blacklist paths, we allow regular expressions to be negated at the outer level.

**Example 1.**  $\neg(e.^i c.^i e)$  indicates that the path must not contain an core switch at the  $(i+1)^{th}$  step. Similarly,  $\neg(e.^i .^i e)$  indicates that the path connecting two edge switches should have a length  $< i+1$ .

Let  $\pi = sw_0 \dots sw_k$  be a path for packet class  $pc$  and let its labeling be  $\Phi(\pi) = a_0 \dots a_k$ . We say that  $\pi$  satisfies a tactic  $R$ ,  $\Phi(\pi) \in L(R)$ , if the following holds:

- $\Phi(\pi) \in L(\neg R)$  iff  $\Phi(\pi) \notin L(R)$ ;
- $\Phi(\pi) \in L(l_{src}.^i .^i l_{dst})$  iff  $k \geq i+1$ ,  $l_{src} = a_0$ , and  $l_{dst} = a_k$ ;
- $\Phi(\pi) \in L(l_{src}.^i l.^i l_{dst})$  iff  $k \geq i+2$ ,  $l_{src} = a_0$ ,  $l_{dst} = a_k$ , and  $a_{i+1} = l$ ;
- $\Phi(\pi) \in L(l_{src}.^i l_1 l_2.^i l_{dst})$  iff  $k \geq i+3$ ,  $l_{src} = a_0$ ,  $l_{dst} = a_k$ ,  $a_{i+1} = l_1$ , and  $a_{i+2} = l_2$ .

In Genesis, operators can specify conjunctions of tactics which adhere to the restricted syntax and the synthesis algorithm is modified

<sup>4</sup> It is a subset of star-free languages [8].

to enforce the tactics. Since tactics are path-related restrictions, operators can also apply tactics selectively to the paths for certain packet classes, and also different tactics for different classes.

**Example 2.** “No Edge” Tactic is a complex form of tactic which ensures that a edge-edge path cannot traverse another edge switch.

It is expressed using conjunctions:  $\neg(e.*e.*e) \equiv \bigwedge_{i=0}^{i=\mu-2} \neg(e.^i e.*e)$

where  $\mu$  is the upper limit on path length.

**Example 3.** “Valley-free” Tactic ensures that a edge-edge path is of the form  $e - a - c - a - e$ . It is expressed as  $\neg(e.^5.*e) \wedge \neg(e.*e.*e)$ .

Paths used in production datacenter networks adhere to both these tactics [12, 24]. Such paths are simple and make networks easy to manage and troubleshoot [?].

### 6.1.2 Modified Synthesis Algorithm with Tactics

In our synthesis algorithm, the reachability propagation constraints (Equation (3)) construct the path from destination to source. We use tactics to prune these constraints, so that the path synthesized satisfies the tactic regular expression.

The tactic set  $\Gamma = \{(R_1, pc_1), \dots, (R_n, pc_n)\}$  specifies that tactic  $R_i$  is applied on packet class  $pc_i$  where  $pc_1, \dots, pc_n \in PC$  and  $R_1, \dots, R_n$  are regular expressions satisfying the restricted tactic syntax. Given a tactic  $R$  applied to packet class  $pc$ , we define  $\Psi_T(R, pc)$  as the additional SMT constraints used for synthesis such that  $(\Psi \wedge \bigwedge_{(R, pc) \in \Gamma} \Psi_T(R, pc))$  is provided as input to the

SMT solver. Note that  $\Psi_T(R, pc)$  is presented as additional SMT constraints only for clarity. In practice, the modified synthesis algorithm will remove constraints for each  $(R, pc) \in \Gamma$ .

**Type 1.** For a tactic  $R$  applied to  $pc$  of the form  $\neg(l_{src}.^i l_{dst})$  which restricts the path to a length  $< i + 1$ :

$$\Psi_T(R, pc) = \forall sw, k \geq i + 1. (sw, pc, k) \notin Reach \quad (18)$$

Thus, we can remove the reachability constraints of Equation (3) for all the tuples  $(sw, pc, k) \notin Reach$  satisfying Equation (18) as they cannot contribute to any path satisfying the tactic.

**Type 2.** For a tactic  $R$  applied to  $pc$  of the form  $\neg(l_{src}.^i l_{dst})$ :

$$\Psi_T(R, pc) = \forall sw. \phi(sw) = l \wedge sw \neq dst \implies (sw, pc, i + 1) \notin Reach \quad (19)$$

The tactic ensures that a switch with label  $l$  cannot be reached in  $i + 1$  steps, except if  $l = l_{dst}$ . In that case, only the destination switch with label  $l$  can be reached in  $i + 1$  steps as the path with labeling  $l_{src}.^i l_{dst}$  satisfies the tactic. If  $l \neq l_{dst}$ , then all switches with label  $l$  cannot be reached in  $i + 1$  steps. For all tuples  $(sw, pc, i + 1) \notin Reach$  satisfying Equation (19), we can remove the reachability constraints of Equation (3).

**Type 3.** For a tactic  $R$  applied to  $pc$  of the form  $\neg(l_{src}.^i l_1 l_2.*l_{dst})$ :

$$\Psi_T(R, pc) = \forall n_1, n_2. \phi(n_1) = l_1 \wedge \phi(n_2) = l_2 \wedge n_2 \neq dst \implies \neg(Reach(n_1, pc, i + 1) \wedge Fwd(n_1, n_2, pc)) \quad (20)$$

This tactic ensures that a switch with label  $l_1$  at  $i + 1$  in the path will not forward the packet to a switch with label  $l_2$  (unless  $n_2$  is the destination). To enforce this, we modify the Equation (3) and remove all  $l_1 \rightarrow l_2$  edges at position  $i + 1$  in the path for which the switch with label  $l_2$  is not the destination.

We state the soundness and completeness theorems for the synthesis algorithm with tactics. Let  $(Fwd, Reach)$  be a model of  $\Psi$  and  $\Pi = \text{paths}(Fwd, Reach)$  be the set of induced paths (from Definition 1).

**Theorem 6.1** (Soundness). For a tactic set  $\Gamma$ , if  $(Fwd, Reach) \models \Psi \wedge \bigwedge_{(R, pc) \in \Gamma} \Psi_T(R, pc)$ , then  $\forall (R, pc) \in \Gamma. \forall (\pi', pc') \in \Pi. pc = pc' \implies \Phi(\pi') \in L(R)$ .

**Theorem 6.2** (Completeness). For a tactic set  $\Gamma$ , if  $\Pi \models \Psi$  and  $\forall (R, pc) \in \Gamma. \forall (\pi', pc') \in \Pi. pc = pc' \implies \Phi(\pi') \in L(R)$  then  $\forall (R, pc) \in \Gamma. (Fwd, Reach) \models \Psi_T(R, pc)$ .

The intuition behind the restricted tactic syntax comes from the structure of the reachability propagation constraints (Equation (3)) which construct the path for a packet class. Each constraint enforces that if a switch is reachable in  $k$  steps in a path, there must be a neighbour switch in the path reachable at  $k - 1$  steps. We design the restricted tactic syntax based on this, providing restrictions on the lengths of the path using  $\neg(l_1.^i l_2)$ , or specifying if we cannot reach certain switches at specific positions in the path using  $\neg(l_1.^i c.^i l_2)$ , or specifying local constraints on neighbours using  $\neg(l_1.^i c_1 c_2.^i l_2)$ . Conversely, the structure of these constraints prevents us from being able to specify unrestricted regular expressions.

**Example 4.** Consider a tactic  $\neg(e.^i aca.*e)$ . To enforce this tactic, we need to have constraints which prevent the path reaching an aggregate switch in  $i + 3$  steps when the path traverses an aggregate and core switch at  $i + 1$  and  $i + 2$  steps respectively. This cannot be specified by modifying the reachability constraints in its current form, because the constraints for reachability for a switch in  $i + 3$  steps only depends on the constraints for reachability in  $i + 2$  steps.

Supporting arbitrary regular expressions requires additional constraints to our synthesis algorithm which leads to increased synthesis times. Thus, we only support a restricted tactic syntax.

Since tactics artificially reduce the search space, the synthesis algorithm with tactics is incomplete (a solution satisfying the tactic may not exist). In practice, tactics can be used to specify restrictions which would not reduce the search space dramatically, but are still useful toward speeding up the synthesis, especially in datacenter topologies which are hierarchical and can be used to specify interesting tactics. Some examples are: an edge to edge path must not traverse another edge, and valley-free routing (*ecae*). One of the biggest advantages of tactics is that it is *policy-agnostic* since it enforces conditions on the path, and can be used in conjunction with the different policies supported by Genesis (isolation, traffic engineering etc.). Thus, we have provided a framework for the development of search strategies based on path properties, and we envision entities providing network management to develop a *repository* of tactics for various network topologies, which can be used by network operators without writing tactics of their own.

## 7. Divide-and-Conquer Synthesis

One of the key challenges to synthesis performance is the number of packet classes synthesized and the policy interactions among them (isolation, capacity, etc.). Since the complexity of finding a forwarding plane configuration is roughly *exponential* in the number of packet classes, the synthesis time shoots up with increasing packet classes. We notice that since datacenter topologies have a dense interconnection of links between layers there can be numerous different forwarding plane configurations as solutions. We propose a heuristic approach which partitions the problem into smaller components to speed up synthesis.

The intuition is as follows. Suppose we have two packet classes  $pc_1, pc_2$  isolated from one another; the standard synthesis algorithm adds constraints for both packet classes to the solver for finding the solution. Instead, we could synthesize  $pc_1$  independently and, after that, find a solution for  $pc_2$  that is isolated from the path obtained for  $pc_1$ . We term this procedure *divide-and-conquer* synthesis, as we divide the problem and try to synthesize them sepa-



rately. However, synthesis of  $pc_2$  can fail because the solution of  $pc_1$  may be such that there is no way to place an isolated path for  $pc_2$  but if they had been synthesized together, a solution might exist. To improve effectiveness, we integrate the divide-and-conquer approach with recovery mechanisms discussed in §7.1.

---

**Pseudocode 2** Divide-and-Conquer Synthesis

---

```

1: procedure DCSYN(P)
2:   if  $size(P) < P_{thres}$  then
3:     Apply normal synthesis on P
4:   else
5:     Partition P into  $P_1$  and  $P_2$ 
6:     if  $interpartition\ edges > E_{thres}$  then
7:       Apply normal synthesis on P
8:     end if
9:      $F = []$  /* Failed solutions */
10:     $attempts = 0$ 
11:    while  $attempts < RA_{max}$  do
12:       $sol_1 = \text{Apply synthesis on } P_1 \text{ such that } sol_1 \notin F$ 
13:      if  $\text{synthesis}(P_1)$  fails then
14:        return DCSyn failure
15:      end if
16:       $sol_2 = \text{Apply synthesis on } P_2 \text{ such that}$ 
17:         $sol_2 + sol_1 \text{ is a solution for } P$ 
18:      if  $\text{synthesis}(P_2)$  fails then
19:         $F.append(unsat-cores(P_2))$ 
20:         $attempts++$ 
21:      else
22:        return DCSyn success
23:      end if
24:    end while
25:    return DCSyn failure
26: end procedure

```

---

We define a policy graph  $P = (R, I)$  where every vertex  $r \in R$  is a packet class for a reachability/waypoint policy and edges denote isolation constraints between packet classes. An edge  $(r_1, r_2) \in I$  means that the paths of  $r_1$  and  $r_2$  are isolated from each other. We assume that there are no capacity policies in the input specifications. Given the policy graph  $P$ , we can synthesize each connected component independently, since packet classes in different connected components are not related by any isolation policy, and therefore are independent of each other.

We describe the heuristic synthesis procedure in Pseudocode 1, which takes as input a connected component of the policy graph. The crux of the procedure is that we partition each connected component  $P$  into two smaller components  $P_1$  and  $P_2$  and do the following: 1) synthesize  $P_1$  and obtain a solution  $S_1$ , and 2) for packet classes of  $P_2$  that are isolated to packet classes in  $P_1$ , add the solution  $S_1$  as a constraint to ensure that the packet classes in  $P_2$  will not share the edges of the respective paths obtained in  $P_1$ .

We use *min-cut* partitioning to partition the policy graph connected component into two such that inter-partition edge count is minimized. The rationale is that since we intend to perform the synthesis of both partitions separately, the partitioning should maximize isolation policies within components and minimize those across components. By maximizing isolation policies during synthesis of the component, the partial solution is more likely to be compatible with a complete solution. However, if the min-cut partitioning produces a component smaller than a threshold size, we perform partitioning of the graph into two *equal sized* partitions and minimizing the cut edges between the partitions. We need to ensure that we don't partition the graph smaller than a threshold, as the partial solutions obtained by synthesis of very small partitions

are more likely to conflict with other packet classes. Genesis performs divide-and-conquer synthesis recursively on the components till we cannot partition the component further.

## 7.1 Solution Recovery

While in the best case divide-and-conquer synthesis will lead to a great increase in performance, we need a recovery mechanism in case we cannot find compatible partial solutions. Many SMT solvers track constraints and return an unsatisfiable core [6] when synthesis fails. Informally, the unsatisfiable core is a set of tracked constraints<sup>5</sup> that describe why there wasn't a feasible solution. This helps us track failed partial solutions. Thus, if synthesis of  $P_2$  fails, the unsatisfiable cores obtained will be the paths of the solution of  $P_1$  which are causing the synthesis of  $P_2$  to fail. When performing synthesis of  $P_1$  again, we therefore ensure that we get *different* paths from the unsatisfiable cores we extracted. Basically, we perform a *solver-guided* enumeration of different solutions of  $P_1$  to find a satisfying solution for  $P_2$  for faster convergence. The solution recovery procedure is described in lines 11–23.

There are drawbacks to performing recovery. Since, recovery is a form of enumeration, in cases where the graph is highly constrained (clique), finding a solution can lead to increased number of enumerations, while synthesis without partitioning would provide a solution faster. Thus, we bound the number of enumerations performed by the recovery mechanism and return failure if we don't obtain a solution.

Divide-and-conquer synthesis with recovery is sound, but it is incomplete as we bound the number of enumerations. The success of this approach is directly related to the size of the components (determined by  $P_{thres}$ ). This is because, by synthesizing more packet classes together, we decrease the conflicts arising between partial solutions. The extreme case is when we do not partition the component at all (normal synthesis), which is complete. Thus, to make the synthesis complete with faster convergence, we perform iterations of divide-and-conquer synthesis, and at each iteration we double the partition threshold  $P_{thres}$  if the previous iteration failed. This scheme tries to balance the trade-off between completeness, which requires larger components, and performance, as synthesis is faster on smaller components. In the extreme case, after  $O(\log P)$  iterations,  $P_{thres} > P$  and the divide-and-conquer synthesis routine cannot partition the graph any further, thus rendering the routine complete.

This synthesis approach is more *effective* when there is a large number of solutions and provides a significant improvement. When the problem is highly constrained and the number of solutions is low, the recovery mechanisms and multiple iterations could lead to a degraded performance. We evaluate the performance improvement of divide-and-conquer synthesis with varying isolation workloads in §8.3. A drawback of the divide-and-conquer approach is that it is difficult to apply to global policies (like traffic engineering) primarily because splitting the input problem isn't easy; development of strategies to speed-up global policies is future work.

## 8. Evaluation

**Genesis Prototype:** We have implemented a full working prototype of Genesis in Python. We have implemented an interpreter for the Genesis Programming Language using PLY [3] and the synthesizer using the SMT solver Z3 [7] and its  $\nu Z$  extension for MaxSMT and linear optimization [5]; this outputs the forwarding rules for the switches, which can be provided as input to a SDN controller (e.g., Floodlight [11]) to install over the network. Genesis uses the Metis graph-partitioning library [16] to perform equi-sized

<sup>5</sup>Not necessarily a minimal set of constraints.

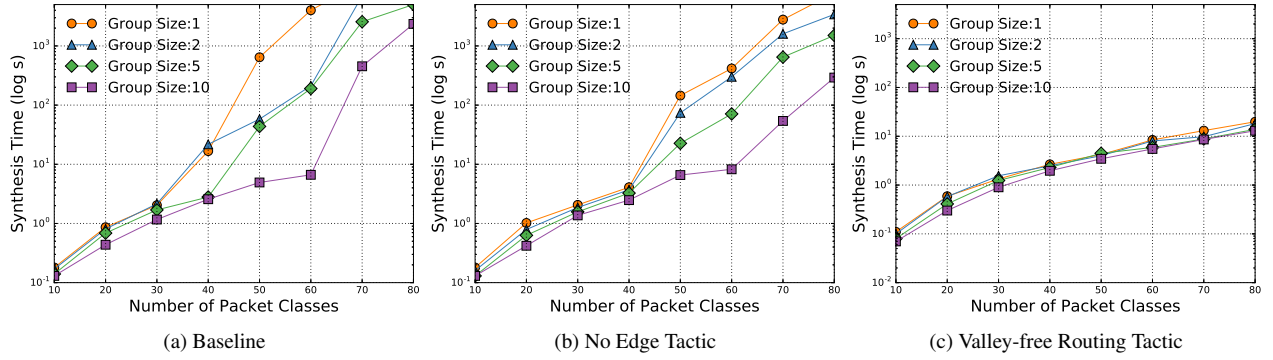


Figure 5: Total synthesis time (log scale) for isolation workloads over range of packet classes and different tenant-group sizes.

partitioning used by divide-and-conquer synthesis. The prototype contains 5K Python lines of code.

In this section, we evaluate Genesis using enterprise-scale multi-tenant data center settings. Specifically, we ask:

- What is the performance of Genesis’s baseline synthesis algorithm? How does the performance vary with size of the network, number and the nature of policies in use? (§8.1)
- How much do tactics help improve Genesis’s performance? Which tactics offer the best improvement? (§8.2)
- To what extent does the divide-and-conquer synthesis improve Genesis’s performance? When does it lead to degraded synthesis times? (§8.3)
- What is the performance of Genesis for traffic engineering and network repair which makes use of SMT with linear optimization objectives, and MaxSMT? (§8.4)

Our experiment settings have a few thousand servers, tens of switches, and hierarchical fat-tree network topologies which reflect a private datacenter. Our experiments are parameterized by: (a) total size of the fat-tree network which we vary between 45 to 180 switches, (b) number of tenants (1-80), and (c) number of packet classes in a tenant (1-10).

Our primary metric of interest is synthesis time, measured in seconds. In measuring this, we focus on the time the Z3 solver takes to solve the constraints<sup>6</sup>. For evaluating the baseline performance, we impose a synthetic limit on the path length  $\mu$  to be 10, which is more than adequate for a fat-tree topology with three levels.

### 8.1 Baseline Synthesis Performance

**Multi-tenant Isolation:** To evaluate the baseline performance of Genesis without tactics, we model a multi-tenant 80 switch topology with tenant-isolation in Figure 5(a). For each workload we have  $n$  tenants with group size  $g$  which is the number of packet classes for each tenant. The x-axis shows the total packet classes  $n * g$ . Packet classes of a tenant are not isolated (and they implement simple reachability within the tenant), while packet classes of different tenants are traffic-isolated. Thus, no two tenants share a link in the same direction, and can never affect each other’s performance. We randomly<sup>7</sup> place endpoints for the tenants’ packet classes, ensuring that not more than 4 tenants share a single edge switch. Operators can aggregate a tenant’s traffic from multiple instances connected to the same switches as a single reachability policy and establish pathways for communication amongst different switches.

For a fixed group size, we observe that, the total synthesis time increases exponentially with number of packet classes. As we

decrease the group size, the synthesis time increases greatly for the same number of packet classes. In this case, the number of tenants increases, making the problem more constrained due to increased number of isolation policies. Group size 1 denotes the extreme case where all flows are isolated with each other.

While we evaluated a multi-tenant isolation setting, there are other scenarios that translate to these workloads. Consider an example where specific flows of tenants require QoS guarantees and these flows must be isolated w.r.t all other flows. This translates to a two-tenant isolation setting. Operators can provide weaker isolation such that two flows must be isolated on only certain “special” links. This is an easier problem to tackle than isolation over all links, and the performance of such scenarios would be better. Failure resiliency uses link-isolation policies which exhibit a similar performance compared to the workloads considered here.

**Effect of Topology Size:** To evaluate Genesis across increasing topology sizes for isolation workloads, we fix the tenant-group size to 5, and for each topology, we maintain the ratio of packet classes to number of edge-aggregate links to 0.25. We choose this metric because if we kept the number of classes constant, as topology sizes increases, it is easier to find isolated paths due to more links. Thus, by keeping the number of packet classes proportional to size of the topology, we maintain the relative difficulty of the workload across topologies. We show the average synthesis time per class with increasing topology sizes in Figure 6 (baseline trace). We are able to synthesize forwarding rules for 12 tenants with group size 5 in a 125 switch topology in 124 seconds (avg. 2 seconds per traffic class). We also observe that average time per flow increases exponentially with larger topologies, thus the synthesis problem is also exponential in number of switches.

**Isolation with Link Capacity Policies:** Figure 7 (baseline trace) shows the average synthesis time per flow for the same setting as above, but additionally, there are 10 low-bandwidth links in the network for which the operator specifies capacity policies (all packet classes have uniform capacity). Since we use LRA for link capacity constraints, we see an increase in average time for synthesis when compared to pure isolation which is completely encoded using SAT.

**Waypoint Policies:** To evaluate Genesis’s performance for ordered sets of waypoints, we fix the number of waypoints (range:1-5) and generate 100 waypoint policies with different sizes and permutations of the ordered waypoint sets for a 80 switch topology. Each policy has edge switches as endpoints and randomly picked core or aggregate switches for waypoints. The synthetic limit  $\mu$  on the path length is increased to 15 and no tactics are used (difficult to devise a tactic for the path satisfying a waypoint policy). The average synthesis time for a waypoint policy is reported in Table 2. We observe that synthesis time increases exponentially with total number of waypoints in a packet class’s policy, owing to the complexity of the problem. Genesis can synthesize rules for a path with 3 total waypoints in less than a second on average. These waypoint poli-

<sup>6</sup> All experiments were conducted using a 32-core Intel-Xeon 2.40GHz CPU machine and 128GB of RAM.

<sup>7</sup> Smarter placement of tenants could speed-up synthesis as tenant endpoints would be located closer to each other. The placement algorithm can be used to develop specialized tactics.

Number of Waypoints	Avg. Synthesis time per Class (s)
1	0.0347
2	0.1384
3	0.9834
4	15.41
5	32.93

Table 2: Average synthesis time per class for waypoint policies with increasing number of waypoints.

cies had either completely ordered or unordered or waypoint sets of sizes 1 and 2.

## 8.2 Tactic Reductions

Using the baseline synthesis approach can result in high synthesis times due to the large solution space of data planes. We demonstrate the improvements from using tactics for isolation workloads with different number of tenants and group sizes on a 80 switch topology.

**“No Edge” Tactic:** Figure 5(b) shows the synthesis time for isolation workloads using the no edge tactic ( $\neg(e.*e.*e)$ ). We achieve a best-case speedup of  $9.5\times$  over baseline synthesis with this tactic. Using this tactic, Genesis can synthesize forwarding rules for 12 tenants with group size 5 in under 200 seconds.

**“Valley-free” Tactic:** For the same isolation workloads as above, we use the tactic  $\neg(e.^5.*e) \wedge \neg(e.*e.*e)$  which ensures *valley-free routing*, that is paths are of the form *eacae*. The results are shown in Figure 5(c). Using this tactic, Genesis synthesizes forwarding rules for each workload in under 20 seconds and can achieve a best-case reduction of  $400\times$  compared to synthesis without tactics.

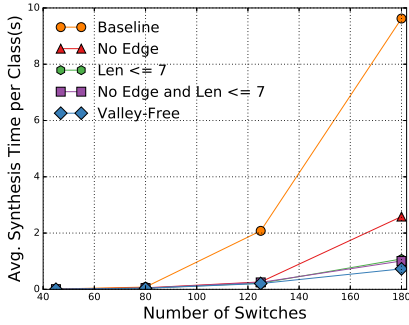


Figure 6: Average synthesis time per packet class versus topology size for isolation workloads w/o different tactics with the ratio of packet classes to number of edge-aggregate links 0.25.

**Effect of Topology Size:** In Figure 6, we evaluate the performance of different tactics for different topology sizes. There is a significant reduction of synthesis time for each tactic when compared to the baseline synthesis. The performance of each tactic is directly related to the reduction of the solution space: a more restrictive tactic has lower synthesis times. Using the no edge tactic and path length  $\leq 7$  tactic, Genesis synthesizes forwarding rules for 20 tenants of group-size 5 in 100 seconds in a 180 switch topology ( $9\times$  speedup over synthesis without tactics).

**Isolation with Link Capacity Policies:** A similar setup with additional link capacity constraints for 10 links is evaluated using the no edge tactic, and we get a best-case  $14\times$  improvement over baseline synthesis. This demonstrates the usefulness of tactics for integrating with different kinds of policies like isolation and link capacities. Tactics can provide a considerable improvement over the baseline performance as illustrated by these experiments, and demonstrate the viability of synthesis approach of Genesis to real-world networks.

Workload Type	Description	Time (s)
minimize-avg-te	100 packet classes	425
	200 packet classes	2002
minimize-max-te	25 packet classes	522
	50 packet classes	4192
Network repair	8 tenants, group size 10, tenant-isolation, 1-switch failure	219

Table 3: Synthesis times for workloads on a 80-node fat-tree topology with different optimization objectives.

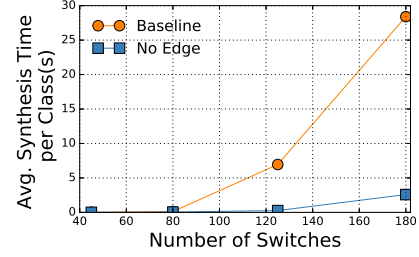


Figure 7: Average synthesis time per packet class versus topology size for isolation workloads with the ratio of packet classes to number of edge-aggregate links 0.25 and 10 low bandwidth links in the topology have capacity policies.

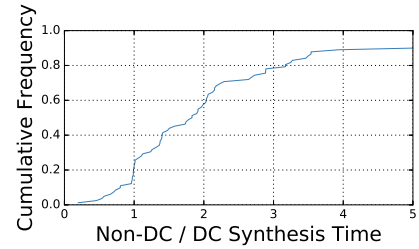


Figure 8: CDF for speedup achieved by divide-and-conquer synthesis.

## 8.3 Divide-and-Conquer (DC) Synthesis Performance

To evaluate the divide-and-conquer (DC) synthesis procedure, we perform 100 runs of DC and non-DC synthesis (with the no edge tactic in both cases) on isolation workloads with varying number of tenants and different group sizes used in §8.1. We compute the speedup (time of non-DC synthesis/time of DC synthesis) and plot its cumulative frequency distribution in Figure 8 to quantify the benefits of DC synthesis. For more than 80% of the workloads, divide-and-conquer offers a better or comparable performance to non-DC synthesis, achieving a speedup of  $2\times$  for nearly 40% of the workloads. For 20% of the workloads, divide-and-conquer performs worse than the non-DC approach, especially for workloads with tenant group size 1. This arises due to greater number of recovery attempts which results in increased time. We can leverage parallelism by running both instances of synthesis, thus giving the best performance, with or without the divide-and-conquer approach.

## 8.4 Synthesis with Optimization Objectives

Table 3 shows the synthesis time for workloads on a 80-node fat-tree topology with different optimization objectives like traffic engineering and network repair. Genesis can do traffic engineering with minimizing average utilization for 200 packet classes in 2000 seconds. However, for minimizing the maximum link utilization, Genesis can only synthesize 50 packet classes in 4000 seconds. For both objectives, the synthesis time increases exponentially with increasing number of packet classes. SMT with optimization objectives is an emerging field of research, and we envision that solvers in the future will become fast to handle larger workloads.

To evaluate the performance of network repair using MaxSMT, we consider a setting with 8 tenants, each with 10 packet classes

(total classes=80), and tenant flows are isolated from one another. Now, we disable the switch with the largest number of rules, and try to find a new configuration from the old one with maximum number of switches unaffected, and the new configuration satisfies the original tenant isolation policies. We can synthesize the minimal repair in near 200 seconds on average.

**Summary.** The key points of our evaluation are:

- For a representative tenant-group size of 10 in a 80 switch fat-tree, the baseline synthesis performance for synthesizing the forwarding rules for 1 to 8 tenants with complete tenant-isolation is in 0.1-2000s;
- Tactics provide considerable speedup over the baseline synthesis. We can synthesize the above workloads in 0.1-300s using the no edge tactic, and under 12s using the valley-free routing tactic;
- Genesis can further benefit from optimistic synthesis, which provides a  $2.0\times$  speed-up over non-optimistic synthesis in 40% of the workloads, in addition to the tactic improvements;
- Adding optimization objectives with SMT for TE and using MaxSMT for network repair is more expensive than synthesis without objectives.

## 9. Related & Future Work

**One Big Switch:** Kang et. al [15] tackle a similar problem of flow policy enforcement. However their end-point policies deal with simple reachability. Their rule placement algorithm takes the path of the flow in the network (called the routing policy) as an input. Genesis supports policies like traffic isolation, for which it cannot determine the path of the flow beforehand. Zhang et. al [28] build on the “one big switch” abstraction [15] to optimize for the specific case of distributed firewall policy enforcement using ILP. PGA [21] provides a graph-level abstraction for specifying network policies like ACLs and middlebox service chaining. However, PGA abstracts the underlying network as “one big switch” and cannot be used to compose policies like tenant isolation or traffic engineering. **Controller synthesis:** Program synthesis has seen limited applications to SDN controllers [19, 27]. These systems synthesize the behavior of individual switches (e.g., learning switches or firewalls); furthermore, these techniques apply to networks operating in a reactive mode (where the first packet of a connection is processed by the controller to determine the actions to employ). Such switch-centric approaches are too constraining and cannot be applied to realize network-wide objectives considered in Genesis.

**Policy languages:** The closest approaches to ours are Merlin [25] and NetGen [23]. In Merlin data planes that adhere to policies expressed using regular expressions are synthesized by first intersecting the topology with the regular expressions appearing in the policies and then encoding reachability in the intersected graph using mixed integer linear programming (ILP). Merlin can support traffic engineering and load distribution. However, in its current iteration, Merlin’s language does not support isolation policies, but we believe that it could be extended to support them. A more prominent difference arises with unordered waypoint policies: expressing a policy including a waypoint set  $W$  of size  $k$  requires a regular expression of size exponential in  $k$  as all the possible permutations of the elements of  $W$  must be considered. This fact clearly impacts the performance of the Merlin’s compiler that would have to generate a mixed ILP with a large number of variables. In Genesis this is not the case as waypoints can be encoded with polynomially many constraints. While this does not affect the theoretical complexity, our compiler does not incur an a-priori exponential blow-up and it rather relies on the power of SMT solvers to guide the search. This is one of the main aspects behind our decision of not using regular expressions to express policies. Genesis uses a restricted form of regular expressions as tactics that leverage the network topology.

While in Merlin regular expressions *increase* the number of constraints generated by the compiler, tactics *decrease* the number of generated constraints therefore speeding up the search. To the best of our knowledge, this is the first use of constraints that leverages the topology structure to simplify the search.

In NetGen, network updates that adhere to policies expressed using regular expressions are synthesized using SMT solvers. Given a specification which mentions the packet classes, the old path, and the new path, NetGen solves the network change problem using an SMT solver. Due to the use of regular expressions NetGen also suffers the limitations we just discussed for Merlin. Interestingly, NetGen uses a specific encoding of regular expressions based on uninterpreted functions that helps reduce the number of constraints. While this encoding is fast when updating a single path, we do not see a way to extend it to our global synthesis setting. A crucial aspect of NetGen is that in its problem formulation each path can be synthesized independently and without affecting the other already synthesized paths. This is not the case when supporting isolation policies: if an old path needs to be moved to satisfy a new policy (e.g., because a link is under maintenance), re-synthesizing such a path can require re-synthesizing other paths.

Synthesis has been also used for generating consistent network updates [17, 29]. But this problem is orthogonal to policy enforcement in Genesis.

**Future directions:** Genesis has support for single-path traffic engineering, however in practice, traffic is split in variable proportions along different paths for effective traffic engineering. One of the future directions of research is to extend the functionality of Genesis to support fine-grained traffic engineering. Also, the performance of SMT solvers with optimization objectives is quite slow, and calls for domain-specific techniques to speed up the synthesis. Also, datacenter networks are highly symmetrical, and this symmetry can be leveraged to speed up synthesis (similar to the work of Plotkin et. al [20] to speed up network verification using symmetry). The main challenges of using symmetry in synthesis is considering two aspects of symmetry: network symmetry and policy symmetry. Also, our treatment of resilience synthesis is preliminary and future work will be geared towards synthesizing resilient forwarding planes incorporating capacity constraints and traffic engineering.

## 10. Conclusion

We presented Genesis, a general and extensible network management system for multi-tenant datacenter networks. It allows rich policies to be specified declaratively. It leverages the formal reasoning foundations of constraint solving together with fast SMT solvers to synthesize data plane configuration from high level policies. This abstracts away the difficult task of programming or configuring individual switches. Genesis incorporates novel ideas to significantly speed up synthesis, leveraging the hierarchical nature of datacenter network topologies and the structure of the interaction between tenants’ policies.

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