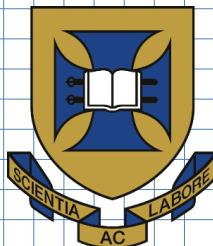


SCOR: Software-defined Constraint Optimal Routing platform for SDN

Siamak Layeghy

April 2018



**THE UNIVERSITY
OF QUEENSLAND
AUSTRALIA**

Outline

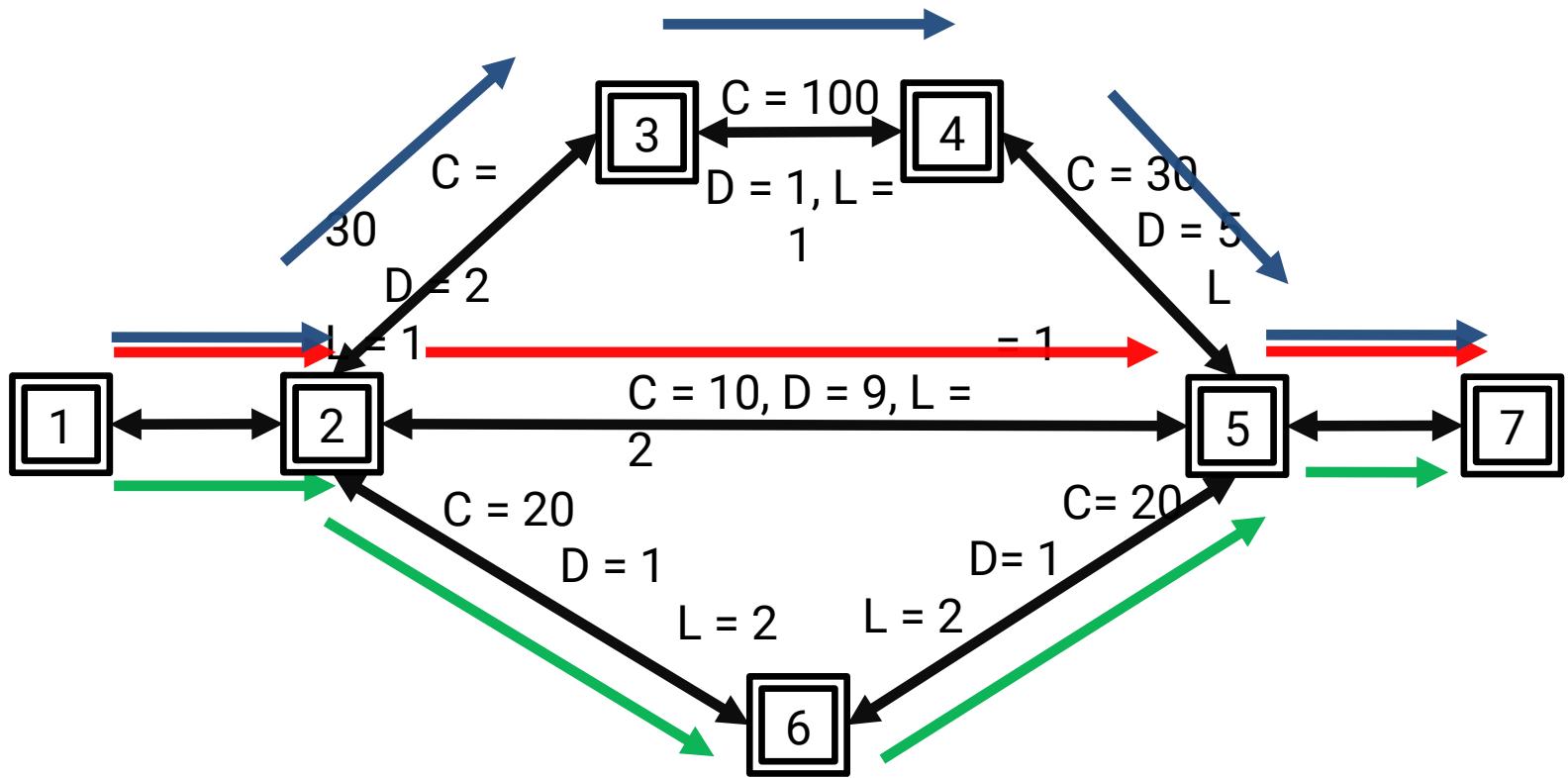
- **Background: QoS Routing**
- **SCOR's Structure**
- **SCOR Models for QoS Routing**
- **Use Cases: ONOS apps**

QoS Routing

Shortest Path Routing

Widest Path Routing (Maximum Bandwidth Routing)

Minimum Delay Path Routing



Routing & QoS Routing

Basic QoS Routing Algorithms

Link-optimization routing

Link-constrained routing

Path-optimization routing

Path-constrained routing

Composite QoS Routing Algorithms

Link-constrained Link-optimization routing

Link-constrained Path-optimization routing

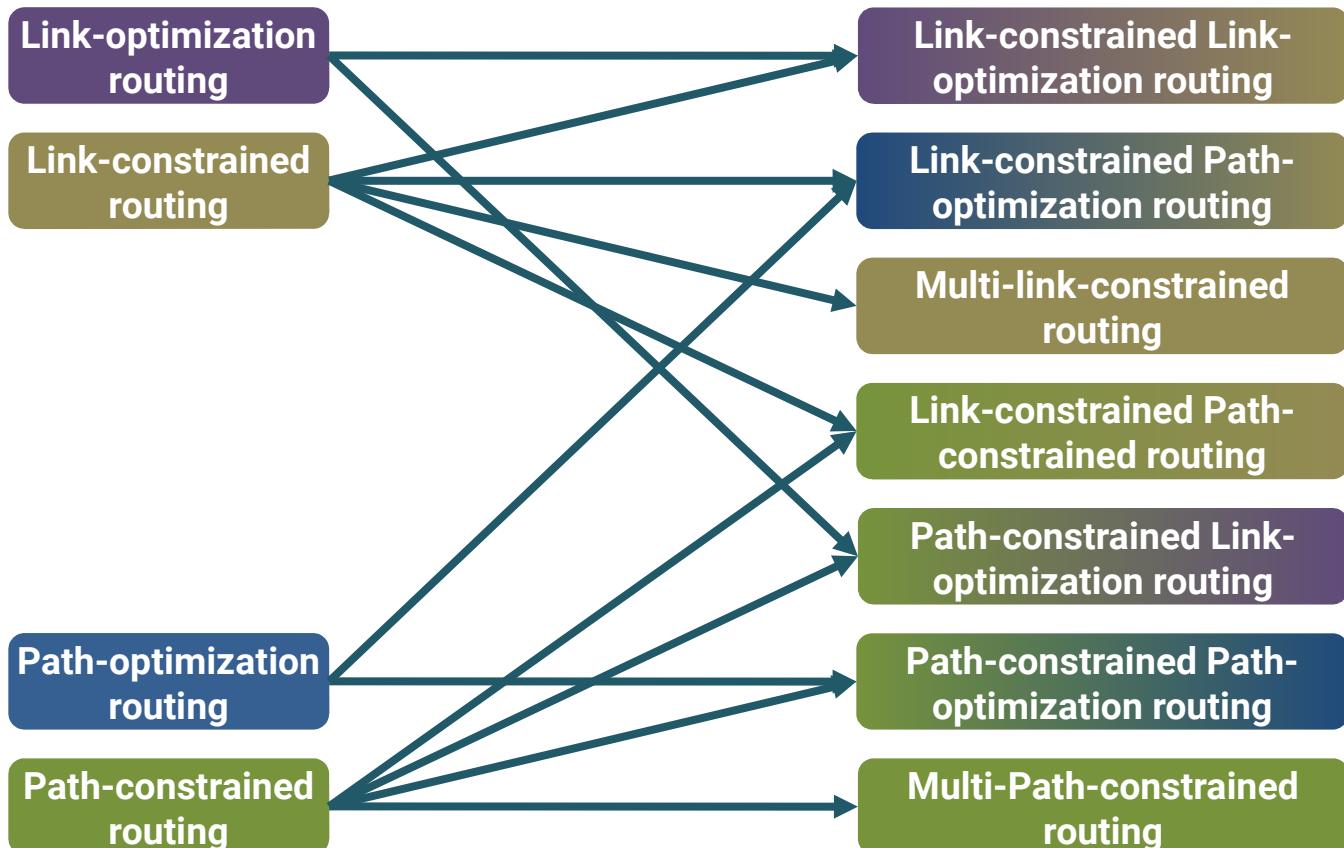
Multi-link-constrained routing

Link-constrained Path-constrained routing

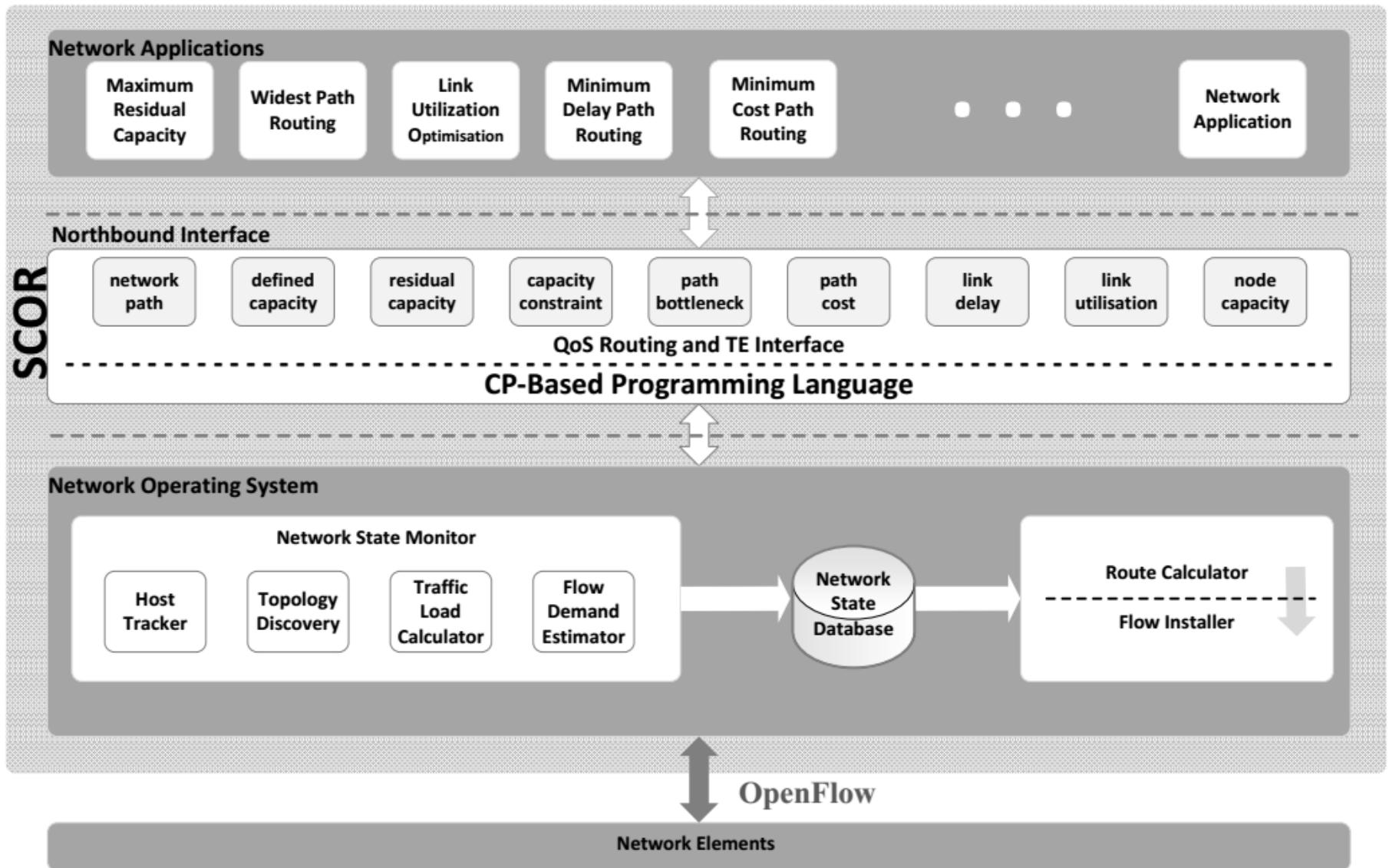
Path-constrained Link-optimization routing

Path-constrained Path-optimization routing

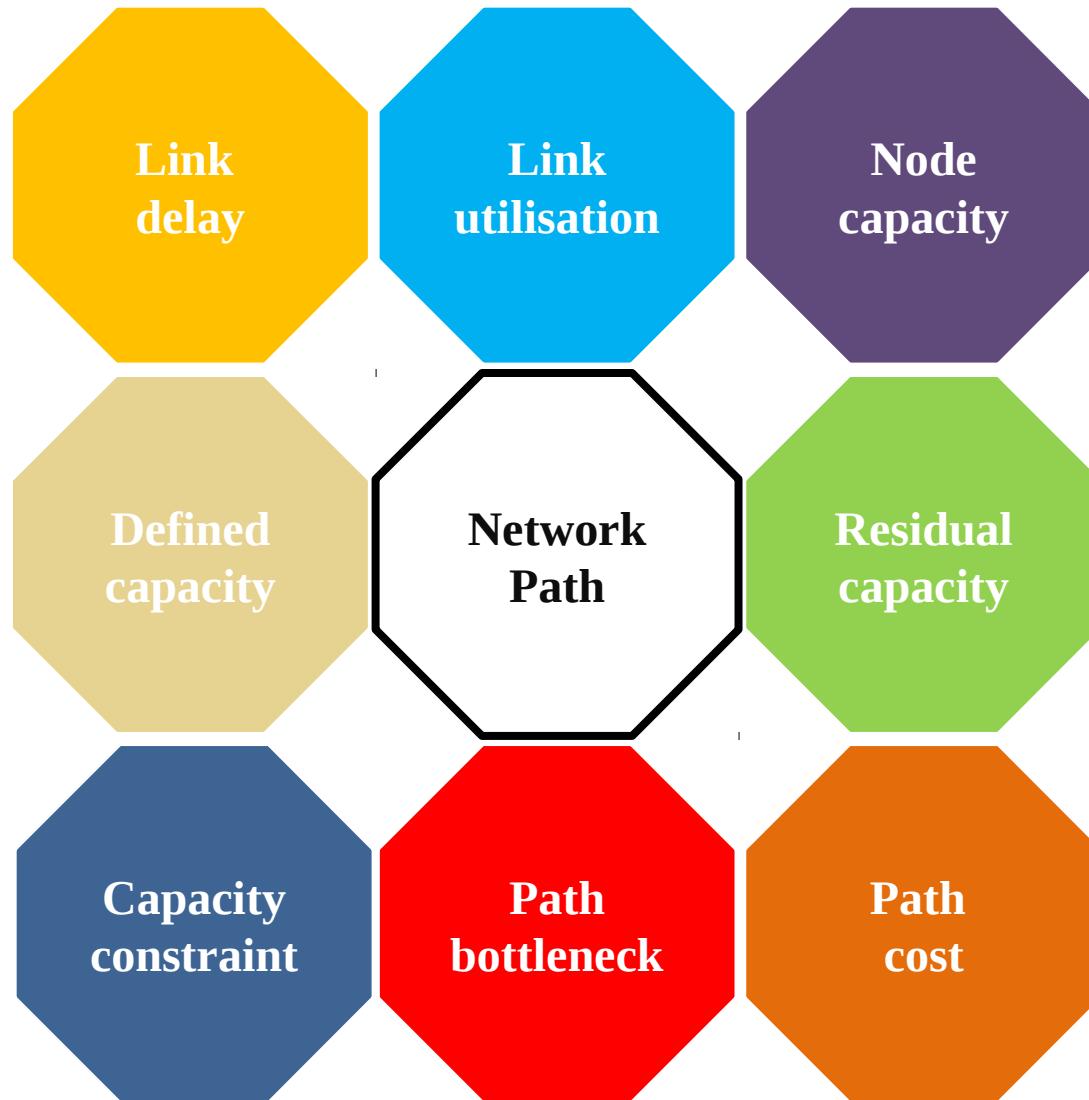
Multi-Path-constrained routing



SDN Routing Framework



SCOR: a New Northbound Interface for Routing



Introducing a New Northbound Interface for QoS Routing

Item	Predicate Name	Implemented Constraint/Defined Value
1	<i>network path</i>	$\sum_{\{v (u,v) \in \mathcal{L}\}} f(u,v) - \sum_{\{v (u,v) \in \mathcal{L}\}} f(v,u) = S_u \quad \forall u \in \mathcal{N}$
2	<i>defined capacity</i>	$c(u,v) \geq c_0 \quad \forall (u,v) \in P_f$
3	<i>residual capacity</i>	$r(u,v) = c(u,v) - f(u,v) \quad \forall (u,v) \in \mathcal{L}$
4	<i>capacity constraint</i>	$f(u,v) \leq c(u,v) \quad \forall (u,v) \in \mathcal{L}$
5	<i>path bottleneck</i>	$c_B[P_f] = \min_{(u,v) \in P_f} \{c(u,v)\}$
6	<i>path cost</i>	$a[P_f] = \sum_{(u,v) \in P_f} a(u,v)f(u,v)$
7	<i>link delay</i>	$D(u,v) = \frac{1}{c(u,v) - f(u,v)} \quad \forall (u,v) \in P_f$
8	<i>link utilisation</i>	$U(u,v) = \frac{f(u,v)}{c(u,v)} \quad \forall (u,v) \in P_f$
9	<i>node capacity</i>	$\sum_{\{v (u,v) \in \mathcal{L}\}} f(u,v) + \sum_{\{v (u,v) \in \mathcal{L}\}} f(v,u) \leq C_u \quad \forall u \in \mathcal{N}$

Introducing a New Northbound Interface for QoS Routing

Predicate 1: Network Path

```
1: forall(i in 1..N)(  
2:     forall(j in 1..F)(  
3:         flow_in_links[i] = sum(k in 1..L)(  
4:             if Links[k,2]=i then LPM[k, j] else 0 endif  
5:         )  
6:         ^  
7:         flow_out_links[i] = sum(k in 1..L)(  
8:             if Links[k,1]=i then LPM[k, j] else 0 endif  
9:         )  
10:        ^  
11:        flow_in_links[i] + (if i = s[j] then 1 else 0 endif) =  
12:        flow_out_links[i] + (bfif i = t[j] then 1 else 0 endif)  
13:        ^  
14:        flow_in_links[i] <= 1  
15:    )  
);
```

Predicate 2: Defined Capacity

```
1: forall(k in 1..L)(  
2:     forall(j in 1..F)(  
3:         if Links[k,4] < Limits[j] then LPM[k,j] = 0  
4:         else true endif  
5:     )  
6: );
```

Predicate 4: Capacity Constraint

```
1: include "Predicate_residual_capacity.mzn";  
2: residual_capacity( LPM, Flows, Links, Residuals );  
3: forall(k in 1..L)(  
4:     Residuals[k]>0  
5: );
```

Predicate 3: residual Capacity

```
1: forall(k in 1..L)(  
2:     if sum(j in 1..F)(LPM[k,j])=0  
3:     then Residuals[k] = Cmax  
4:     else Residuals[k] = Links[k,4] - sum(j in 1..F)(Flows[j] × LPM[k,j])  
5:     endif  
6: );
```

Introducing a New Northbound Interface for QoS Routing

Predicate 6: path bottleneck

```
1: forall(j in 1..F)(  
2:   forall(k in 1..L)(  
3:     Width[k] = Links[k,4] × LPM[k,j]  
4:     )  
5:     Bandwidth[k] = Width[k]  
6:     )  
7:     Bottleneck[j] = min(Bandwidth)  
8: );
```

Predicate 5: path cost

```
1: forall(j in 1..F)(  
2:   Total_Cost[j] =  
3:     sum(k in 1..L)(  
4:       Links[k,3] × LPM[k,j] × Flows[j]  
5:     )  
6: );
```

Predicate 7: link delay

```
1: include "Predicate_capacity_constraint.mzn";  
2: capacity_constraint( LPM, Flows, Links, Residuals );  
3: forall(k in L)(  
4:   if sum(j in 1..F)(LPM[k,j])=0  
5:   then Delay[k] = 0  
6:   else Delay[k] = 1 / Residuals[k]  
7:   endif  
8: );
```

Introducing a New Northbound Interface for QoS Routing

Predicate 9: node capacity

```
1: forall(i in 1..N)(  
2:     node_flow_in[i] = sum(k in 1..L, j in 1..F)(  
3:         if Links[k,2]=i then LPM[k, j] × Flows[j]  
4:         else 0  
5:         endif  
6:     )  
7:     ^  
8:     node_flow_out[i] = sum(k in 1..L, j in 1..F)(  
9:         if Links[k,1]=i then LPM[k, j] × Flows[j]  
10:        else 0  
11:       endif  
12:   )  
13:   ^  
14:   node_flow_in[i] + node_flow_out[i] <= Node_Capacity  
15: );
```

Predicate 8: link utilisation

```
1: forall(k in 1..L)(  
2:     Link_Utilisation[k] = (  
3:         sum(j in 1..F)(  
4:             Flows[j] × LPM[k,j]  
5:         )/ Link[k,4]  
6:         ) × 100  
7:     );
```

Modelling various QoS Routing Algorithms in SDN

Algorithm	Metrics	Type
Shortest Path	Hop Count	Basic
Widest Path	Bandwidth	Basic
Minimum-Delay Path	Transmission Delay	Basic
Minimum-Loss Path	Packet Loss	Basic
Constrained-Delay Path	Transmission Delay	Basic
Constrained-Bandwidth Path	Bandwidth	Basic
Constrained-Bandwidth -Minimum-Delay Path	Bandwidth, Transmission Delay	Multi-constraint
Maximum-Residual-Capacity Path	Bandwidth	Complex
Constrained-Residual-Capacity Path	Bandwidth	Complex
Minimum-Link-Utilisation Path	Bandwidth/total delay	Complex
Constrained-Link-Utilisation Path	Bandwidth/total delay	Complex
Minimum-Congestion Path	Bandwidth/total delay	Complex
Constrained-Congestion Path	Bandwidth/total delay	Complex

Least Cost Path Routing

$$\begin{aligned} & \text{minimise} && \sum_{(u,v) \in \mathcal{L}} a(u,v)f(u,v) \\ & && \sum_{\{v|(u,v) \in \mathcal{L}\}} f(u,v) - \sum_{\{v|(u,v) \in \mathcal{L}\}} f(v,u) = S_u, \quad \forall u \in \mathcal{N} \end{aligned}$$

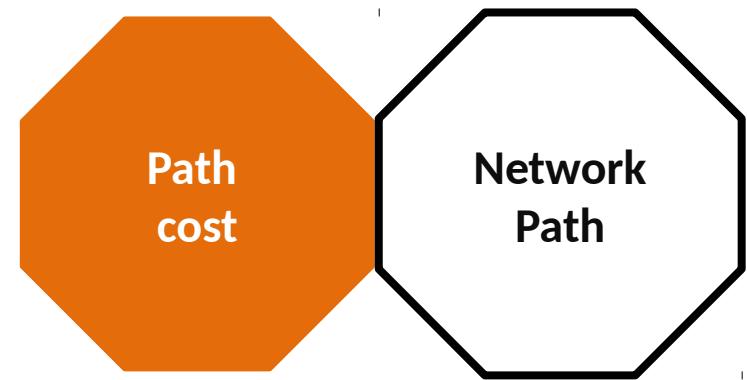
$$f(u, v) = f_1(u, v) = d_1 x_1(u, v)$$

$$\begin{aligned} & \text{minimise} && \sum_{(u,v) \in \mathcal{L}} a(u,v)x_1(u,v) \\ & && \sum_{\{v|(u,v) \in \mathcal{L}\}} x_1(u,v) - \sum_{\{v|(u,v) \in \mathcal{L}\}} x_1(u,v) = S_u/d_1, \quad \forall u \in \mathcal{N} \end{aligned}$$

SCOR Model: Least Cost Path Routing

Model 1: Least Cost Path Routing in SCOR

```
% Include item  
1 : include "Predicate_network_path.mzn";  
2 : include "Predicate_path_cost.mzn";  
  
% Parameters  
3 : array[int,4] of int : Links;  
4 : int : L = max(index_set_1of2(Links));  
5 : array[int] of int : Nodes;  
6 : int : Flows;  
7 : int : s;  
8 : int : t;  
  
% Decision Variables  
9 : var int : Cost;  
10 : array[1..L, 1] of var 0..1 : LPM;  
  
% Constraints item  
11 : constraint network_path(LPM, Links, Nodes, s, t);  
12 : constraint path_cost(LPM, Links, Cost, Flows);  
  
% Solve item  
13 : solve minimize Cost;
```



Least Cost Path Constrained Capacity Routing

$$\text{minimise} \quad \sum_{(u,v) \in \mathcal{L}} a(u,v)f(u,v)$$

$$\sum_{\{v|(u,v) \in \mathcal{L}\}} f(u,v) - \sum_{\{v|(u,v) \in \mathcal{L}\}} f(v,u) = S_u, \quad \forall u \in \mathcal{N}$$

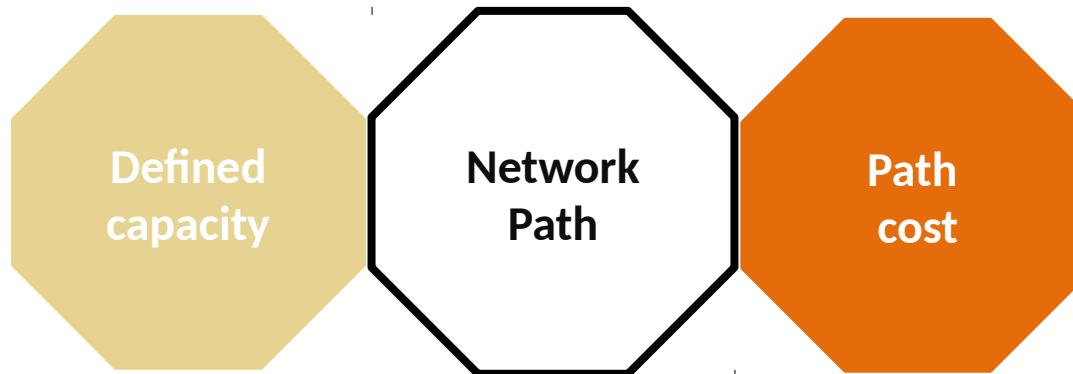
$$f(u,v) = f_1(u,v) = d_1 x_1(u,v)$$

$$\text{minimise} \quad \sum_{(u,v) \in \mathcal{L}} a(u,v)x_1(u,v)$$

$$\sum_{\{v|(u,v) \in \mathcal{L}\}} x_1(u,v) - \sum_{\{v|(u,v) \in \mathcal{L}\}} x_1(u,v) = S_u/d_1, \quad \forall u \in \mathcal{N}$$

$$x_1(u,v) \leq c(u,v)/d_1 \quad \forall (u,v) \in \mathcal{L}$$

SCOR Model: Least Cost Path Constrained Capacity Routing



Model 2: Least Cost Path with Defined Capacity Routing in SCOR

```
:
% Include item
1: include "Predicate defined capacity.mzn";
:
% Parameters
2: int: Limit;
:
% Constraints item
3: constraint defined_capacity(LPM, Links, Flows, Limit);
:
```

Maximum Residual Capacity Path Routing

maximise{ \mathcal{Z} }

$$\mathcal{Z} \leq r(u, v) \quad \forall (u, v) \in \mathcal{L}$$

$$\sum_{\{v|(u,v) \in \mathcal{L}\}} x_j(u, v) - \sum_{\{v|(u,v) \in \mathcal{L}\}} x_j(v, u) = S_u^j/d_j \quad \forall u \in \mathcal{N}, j = 1..F$$

$$\sum_{j=1..F} d_j x_j(u, v) \leq c(u, v) \quad \forall (u, v) \in \mathcal{L}$$

$$r(u, v) = c(u, v) - f(u, v) \quad \forall (u, v) \in \mathcal{L}$$

$$f(u, v) = \sum_{j=1..F} f_j(u, v) = \sum_{j=1..F} d_j x_j(u, v) \quad \forall (u, v) \in \mathcal{L}$$

$$x_j(u, v) = \begin{cases} 0 & (u, v) \notin P_j \\ 1 & (u, v) \in P_j \end{cases}$$

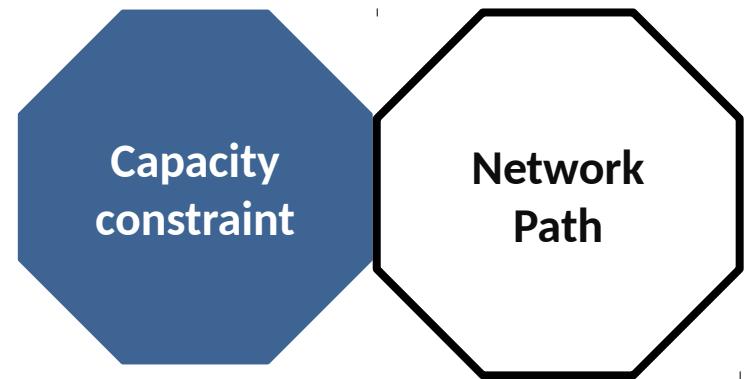
$$S_u^j/d_j = \begin{cases} 1 & u = s_j, \\ -1 & u = t_j, \\ 0 & otherwise \end{cases} \quad u \in \mathcal{N}, j = 1..F$$

SCOR Model: Maximum Residual Capacity

Path Routing

Model 3: Maximum Residual Capacity Routing in SCOR

```
% Include item  
1 : include "Predicate network path.mzn";  
2 : include "Predicate capacity constraint.mzn";  
  
% Parameters  
3 : array[int, int] of int : Links;  
4 : int : L = max(index_set_1of2(Links));  
5 : array[int] of int : Nodes;  
6 : array[int] of int : Flows;  
7 : int : F = max(index_set(Flows));  
8 : array[1..F] of int : s;  
9 : array[1..F] of int : t;  
  
% Decision Variables  
10 : array[1..L, 1..F] of var 0..1 : LPM;  
11 : array[1..L] of var int : Residuals;  
  
% Constraints item  
12 : constraint network_path(LPM, Links, Nodes, s, t);  
13 : constraint capacity_constraint(LPM, Links, Flows, Residuals);  
  
% Solve item  
14 : solve maximize min(Residuals);
```



MRCPR in Procedural Programming

```
str(len(G.node.keys()))+"," +str(len(edges))+") w/ error " + str(error)+ ":" +  
str(shortestPathComputations)  
    for node in G.node.iterkeys():  
        print node, G.edge[node]  
    return shortestPathComputations, count  
  
  
def get_beta_hat(edges, commodities, error=GLOBAL_ERROR, karakosta=True):  
    spc, beta_hat = maximum_concurrent_flow(edges, commodities, error=1., returnBeta=True,  
karakosta=karakosta,  
                                                scale_beta=False)  
    return beta_hat, spc  
  
  
def two_approx(edges, commodities, error=GLOBAL_ERROR, karakosta=True):  
    beta_hat, spc = get_beta_hat(edges, commodities, error=1., returnBeta=True, karakosta=karakosta)  
    scale_demands(commodities, beta_hat / 2.)  
    return maximum_concurrent_flow(edges, commodities, error=error, karakosta=karakosta,  
shortestPathComputations=spc)  
  
def multi_route(edges, commodities, error=GLOBAL_ERROR, scale_beta=True, karakosta=False):  
    beta_hat, spc = get_beta_hat(edges, commodities, error=1.)  
    return maximum_concurrent_flow(edges, commodities, error=error, karakosta=karakosta,  
shortestPathComputations=spc,  
  
                                scale_beta=scale_beta, multi_route=True, beta_hat=beta_hat)  
  
    commodityTable[commodity] = 0  
    for head in G.edge.iterkeys():  
        for tail, edge_dict in G.edge[head].iteritems():  
            for commodity in commodities:  
                if tail == commodity.sink:  
  
commodityTable[commodity]+=edge_dict[FLOW_ATTRIBUTE]/commodity.demand  
print "Lambda is " + str( min([x for x in commodityTable.itervalues()]))  
print "OBJECTIVE: ", calculate_dual_objective(G)  
print "SPC-"+str(karakosta)+"-"+str(twoApprox)+" for G("
```

Half-duplex Maximum Residual Capacity Path Routing

maximise $\{\mathcal{Z}\}$

$$\mathcal{Z} \leq r(u, v) \quad \forall (u, v) \in \mathcal{L}$$

$$\sum_{\{v|(u,v) \in \mathcal{L}\}} x_j(u, v) - \sum_{\{v|(u,v) \in \mathcal{L}\}} x_j(v, u) = S_u^j/d_j \quad \forall u \in \mathcal{N}, j = 1..F$$

$$\sum_{j=1..F} d_j x_j(u, v) \leq c(u, v) \quad \forall (u, v) \in \mathcal{L}$$

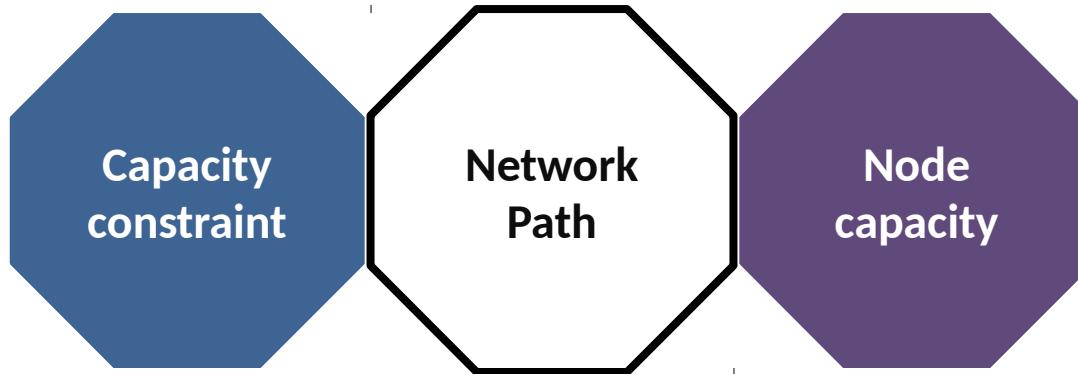
$$r(u, v) = c(u, v) - f(u, v) \quad \forall (u, v) \in \mathcal{L}$$

$$f(u, v) = \sum_{j=1..F} f_j(u, v) = \sum_{j=1..F} d_j x_j(u, v) \quad \forall (u, v) \in \mathcal{L}$$

$$x_j(u, v) = \begin{cases} 0 & (u, v) \notin P_j \\ 1 & (u, v) \in P_j \end{cases} \quad \sum_{\{v|(u,v) \in \mathcal{L}\}} f(u, v) + \sum_{\{v|(u,v) \in \mathcal{L}\}} f(v, u) \leq C_u \quad \forall u \in \mathcal{N}$$

$$S_u^j/d_j = \begin{cases} 1 & u = s_j, \\ -1 & u = t_j, \\ 0 & otherwise \end{cases} \quad u \in \mathcal{N}, j = 1..F$$

SCOR Model: Half-duplex Maximum Residual Capacity Path Routing



Model 4: Half-duplex Maximum Residual Capacity Routing in SCOR

(Only lines not included in maximum residual capacity routing)

⋮

% Include item

1 : **include** "Predicate node capacity.mzn";

⋮

% Parameters

2 : **array[int] of int** : *Nodes_Capacities*;

⋮

% Constraints item

3 : **constraint** *node_capacity(LPM, Links, Flows, Nodes_Capacities)*;

⋮

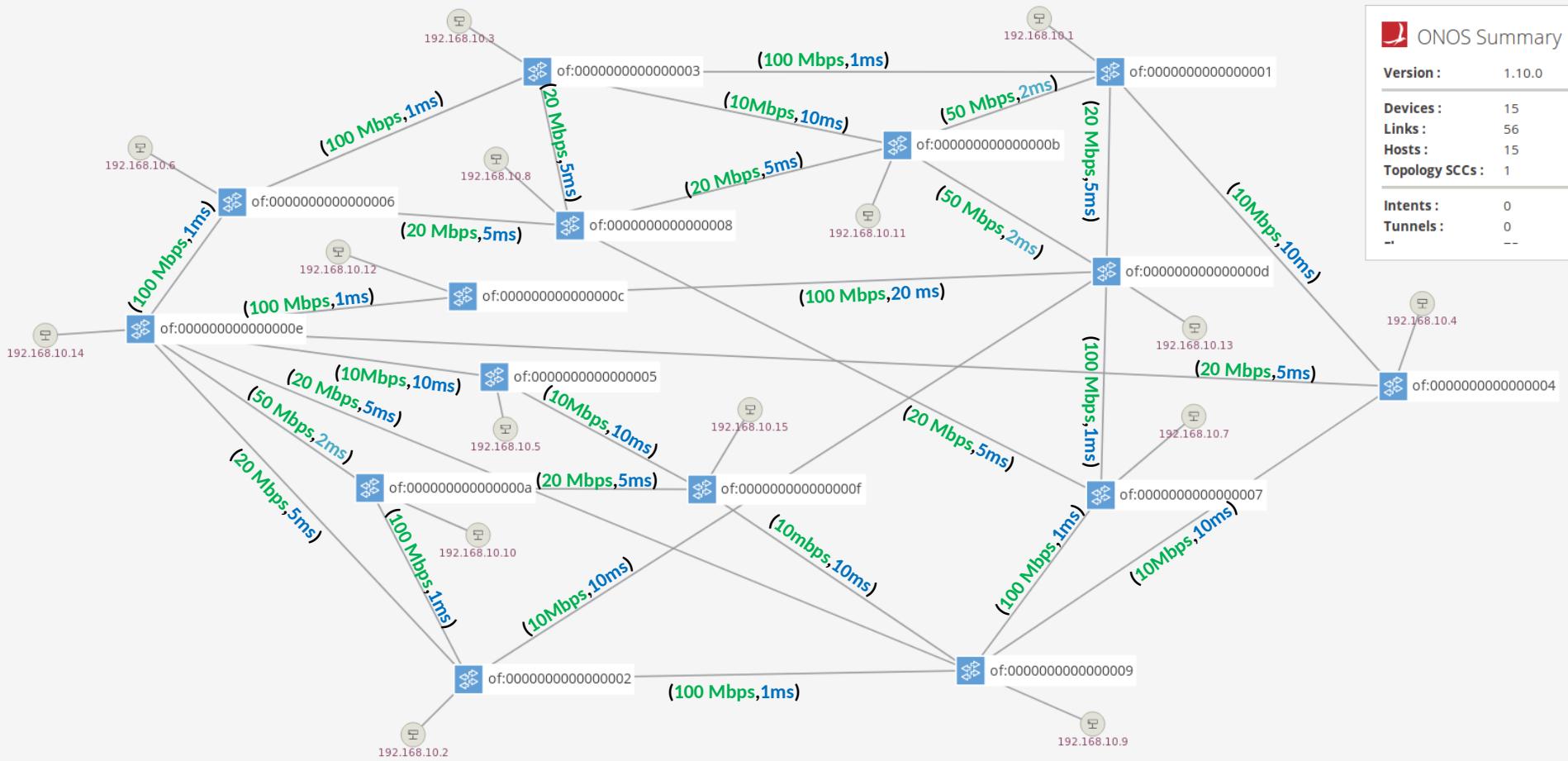
Conciseness & Completeness

QoS Routing Problem	SCOR Predicates	#L
Shortest Path	path cost	3
Widest Path [11]	path bottleneck	3
Bandwidth-Guaranteed [13]	capacity constraint	3
Bandwidth-Constrained [11]	defined capacity	3
Minimum-Loss [14]	path cost	3
Minimum-Delay [14]	path cost	3
Minimum-Delay [64]	delay	3
Delay-Constrained [11]	path cost	3
Least-Cost [11]	path cost	3
Maximum Residual Capacity [12]	capacity constraint	3
Minimum Link Utilisation [16]	link utilisation	4
Delay-Constrained Least-Cost [11]	path cost $\times 2$	4
Delay-Delay Jitter-Constrained [11]	path cost $\times 2$	4
Bandwidth-Delay-Constrained [11]	defined capacity path cost	4
Bandwidth-Constrained Least-Delay [11]	defined capacity path cost	4
Minimum-Cost Bandwidth-Constrained [15]	capacity constraint path cost	4
Delay-Constrained Bandwidth-Optimised [11]	path cost path bottleneck	4
Widest Shortest Path [17]	path cost path bottleneck	6
Shortest Widest Path [17]	path cost path bottleneck	6

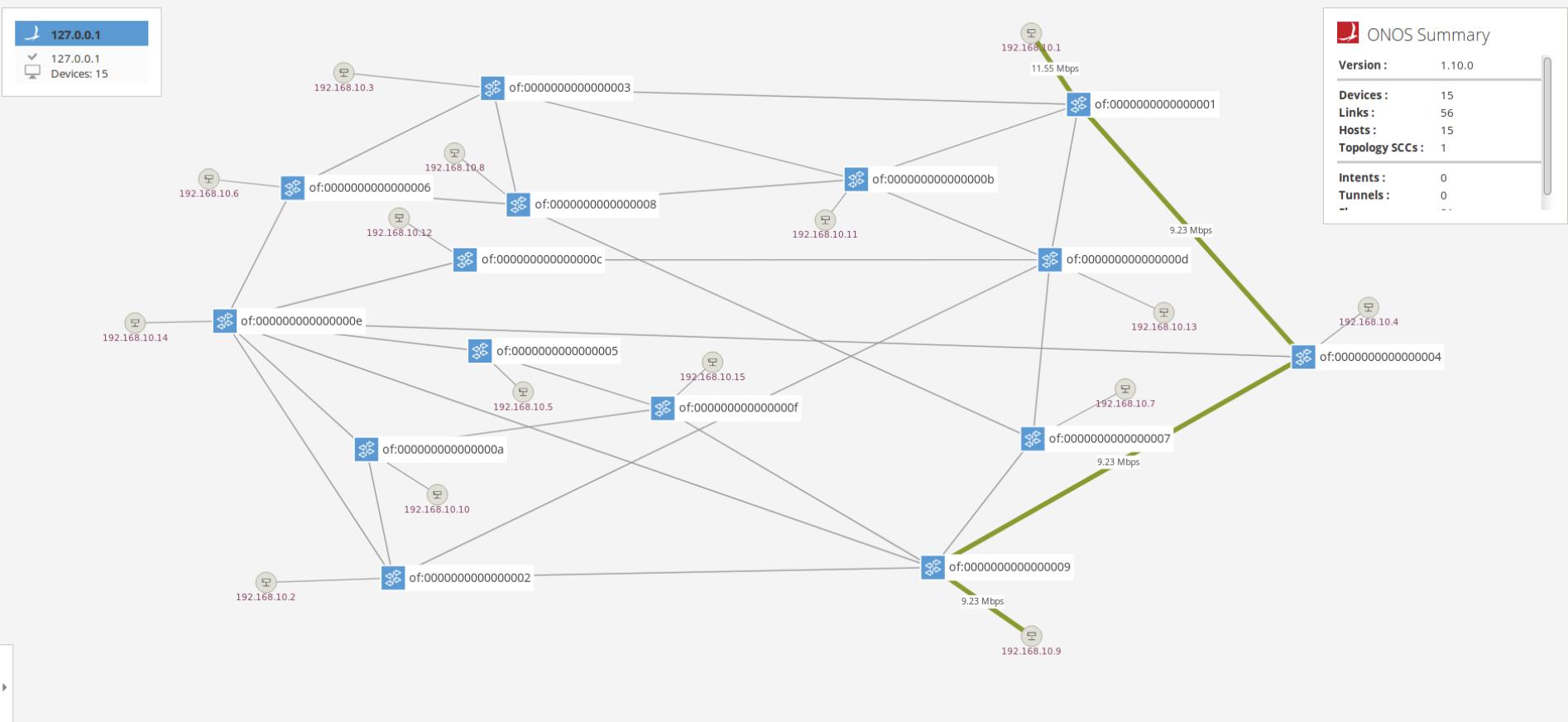
Real-World Use Cases: ONOS Apps

- **Max-BW Routing**
- **Min-Delay Routing**
- **Max-BW Constrained Delay Routing**
- **Min-Delay BW Constrained Routing**
- **Minimizing Max Link Utilization**
- **Maximizing Min Residual Capacity**

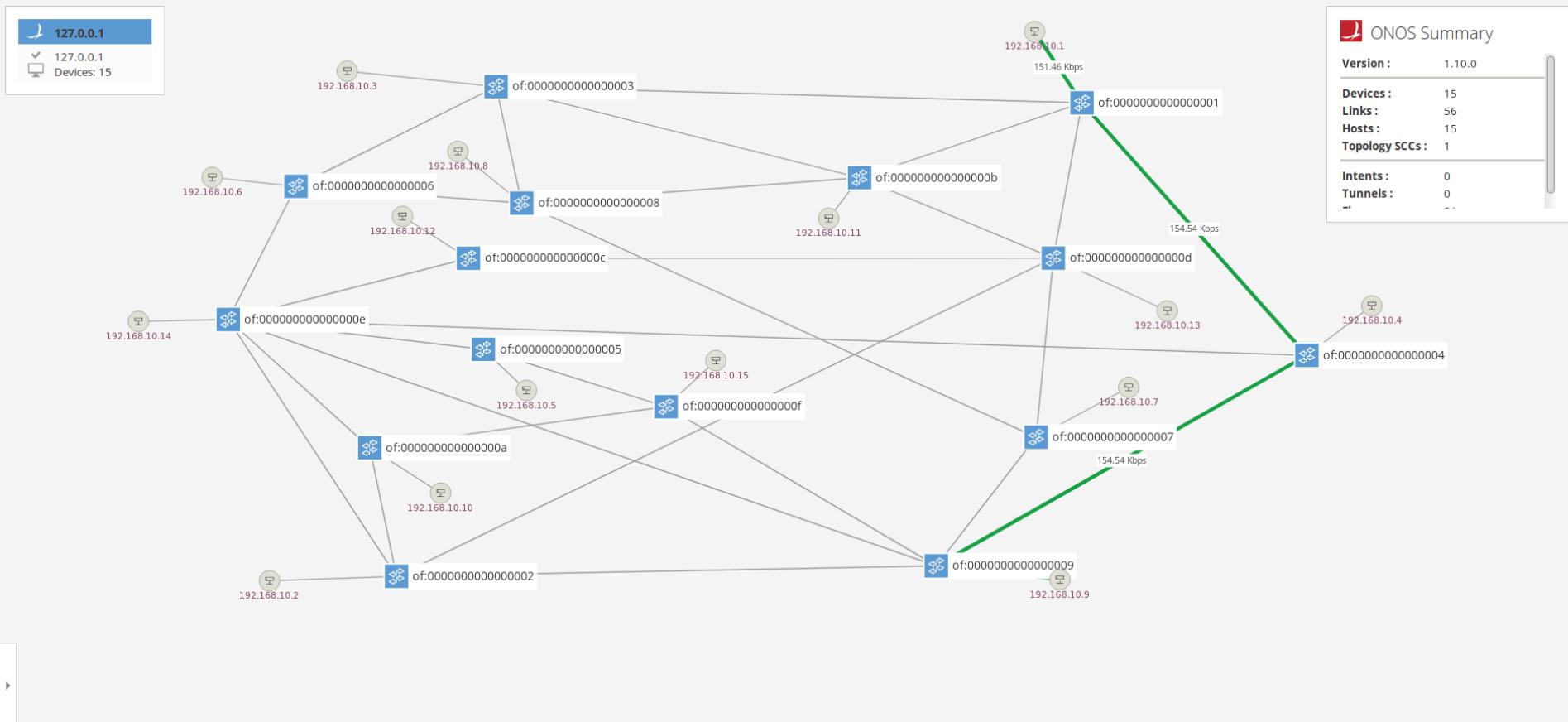
Real-World Use Cases: Scenario



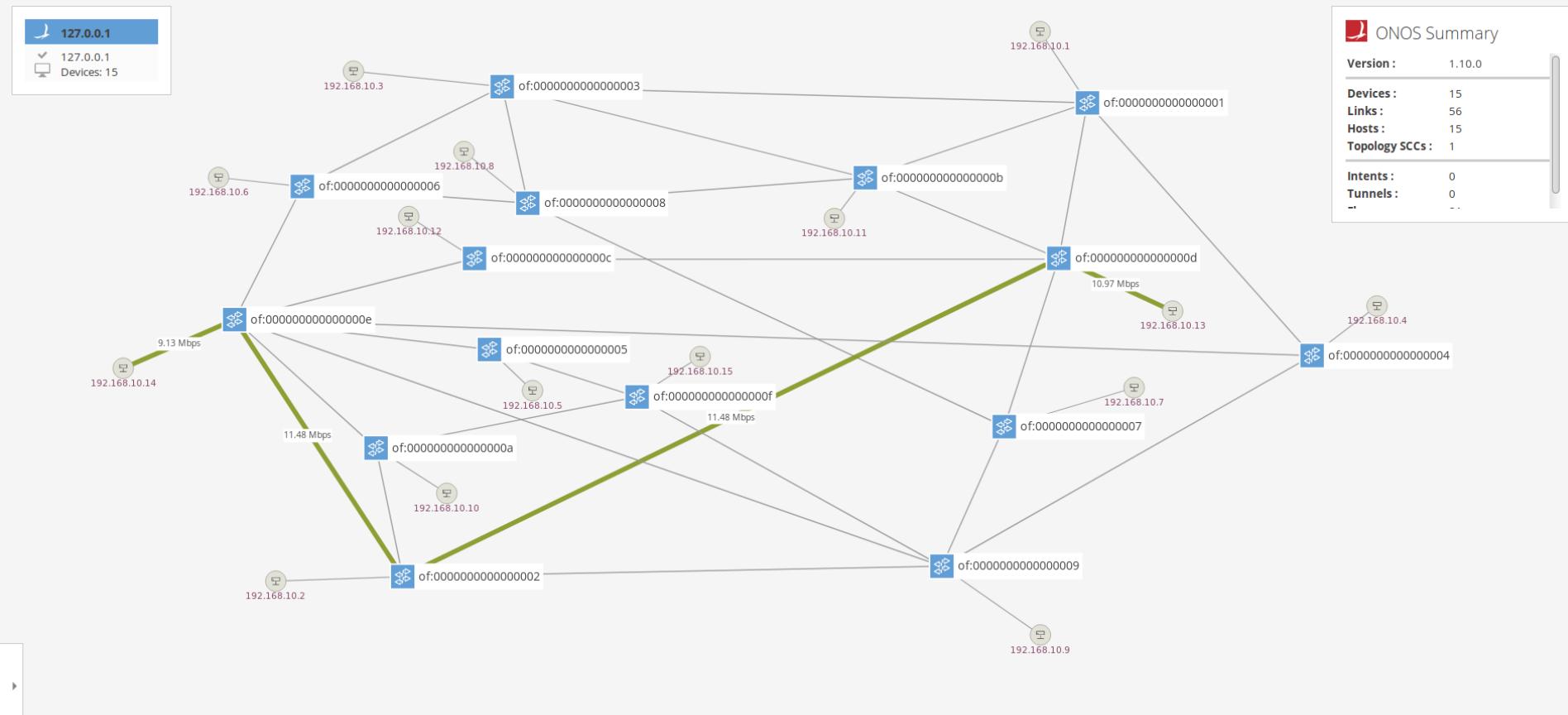
Maximum-Bandwidth Routing



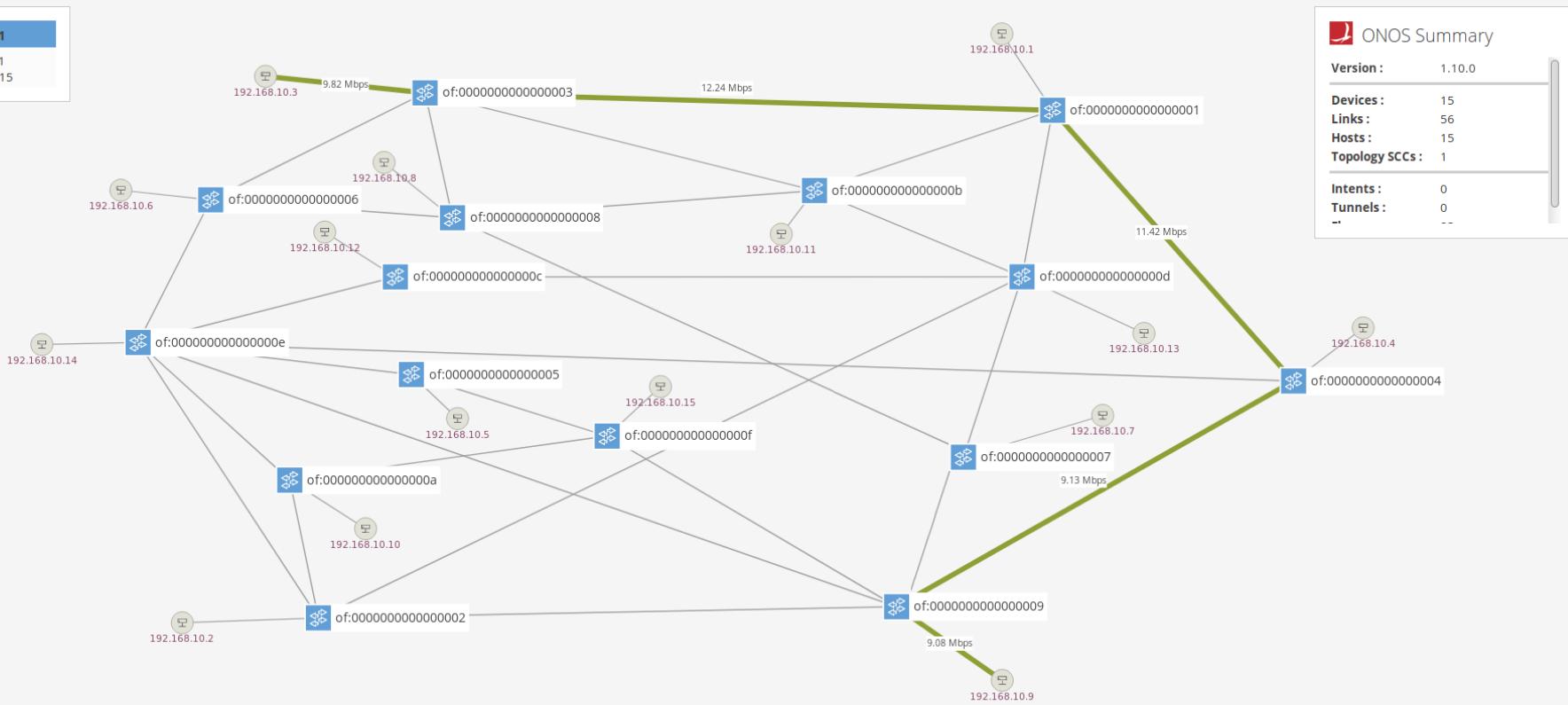
Minimum-Delay Routing



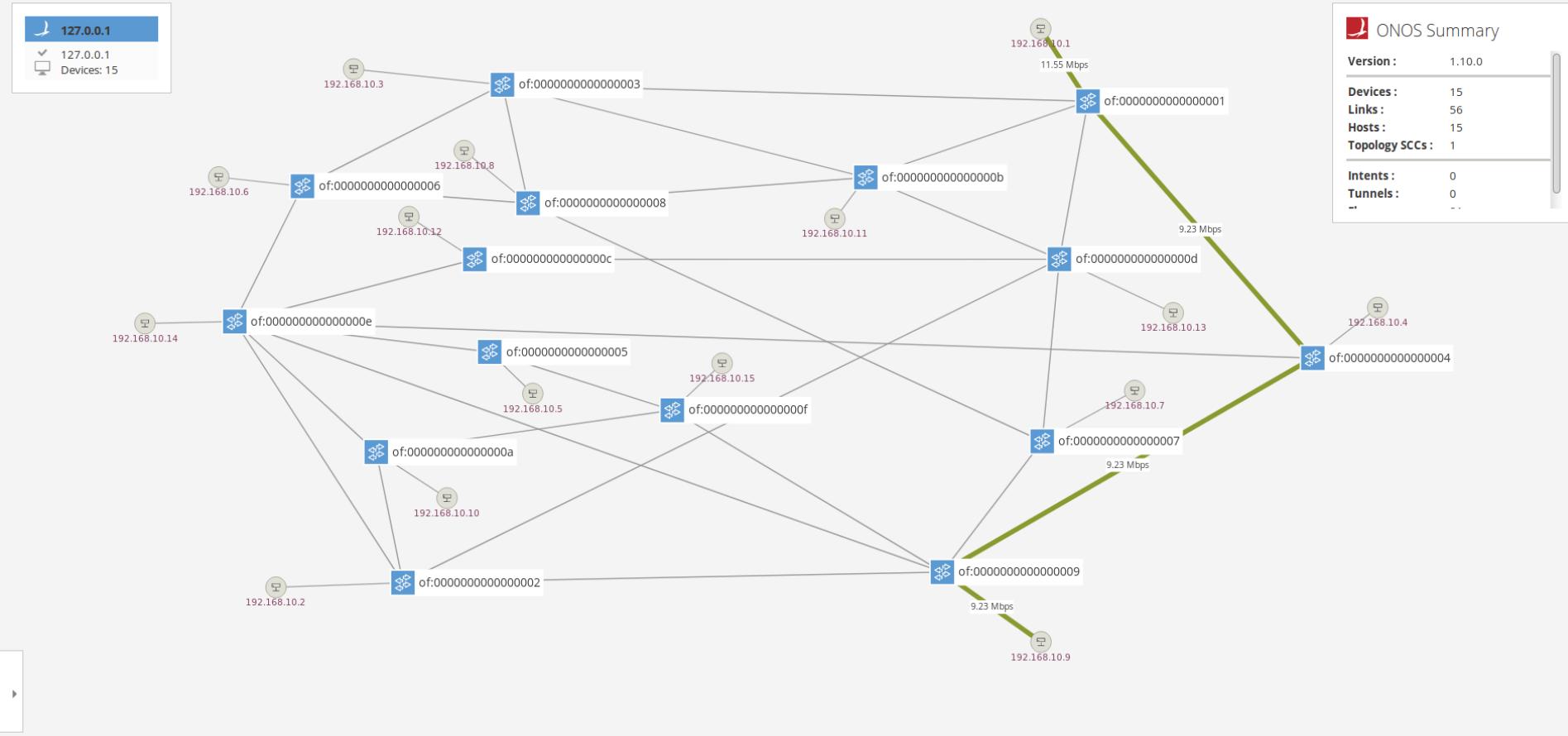
Max-BW Constrained Delay Routing



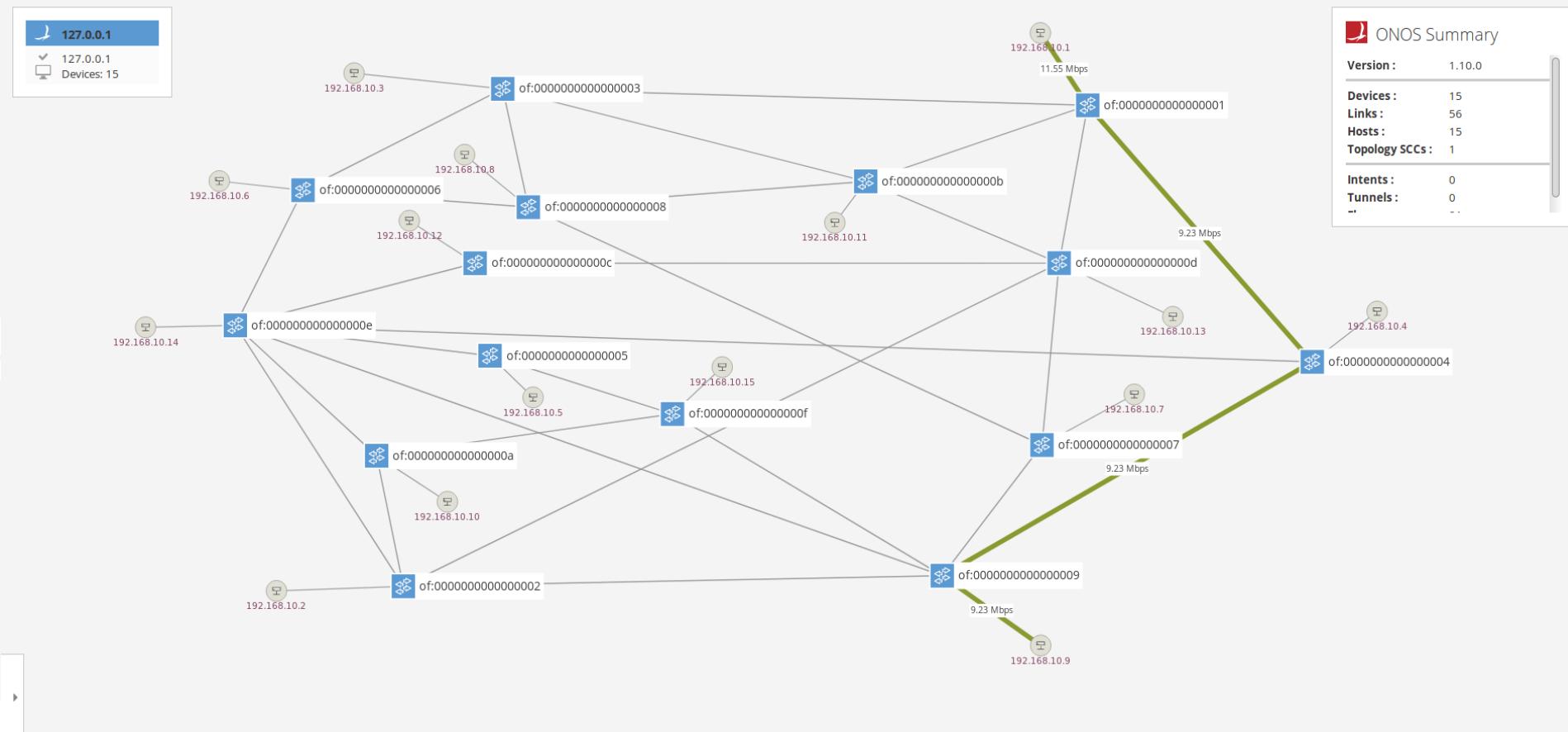
Min-Delay Constrained BW Routing



Minimizing Max Link Utilization



Maximizing Min Residual Capacity



Q & A
Thank You