

# Adaptive Robust Traffic Engineering in Software Defined Networks

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# Traffic Engineering and Software-Defined Networks

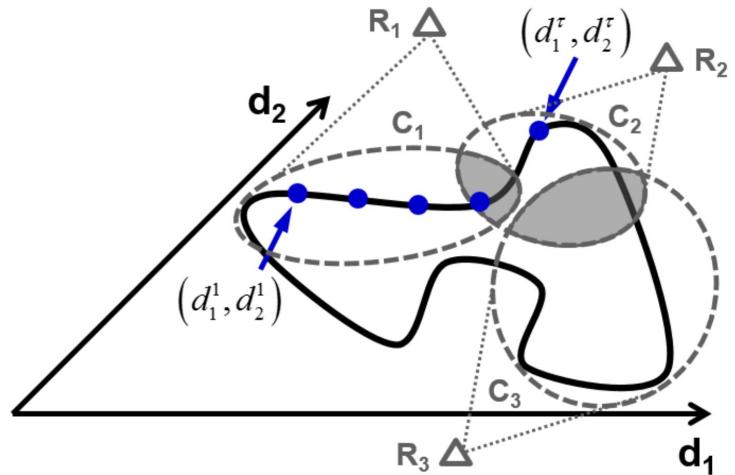
- T.E. optimizes network configuration according to traffic conditions
  - Traffic Matrix (TM)
  - Maximum Link Utilization (MLU)
- Dynamicity of traffic
  - ordinary daily fluctuations
  - unpredictable events (congestion, failures, ...)
- Software-Defined Networks
  - global view of the network status
  - traffic monitoring
  - online traffic optimization

# How to cope with traffic dynamicity?

- Static TE
  - stable routing configuration
  - low optimality
- Dynamic TE
  - multiple routing configurations
  - optimal routing
  - traffic monitoring latency and processing overhead
  - routing instability (consistent update)
- Semi-static TE

# Our contribution

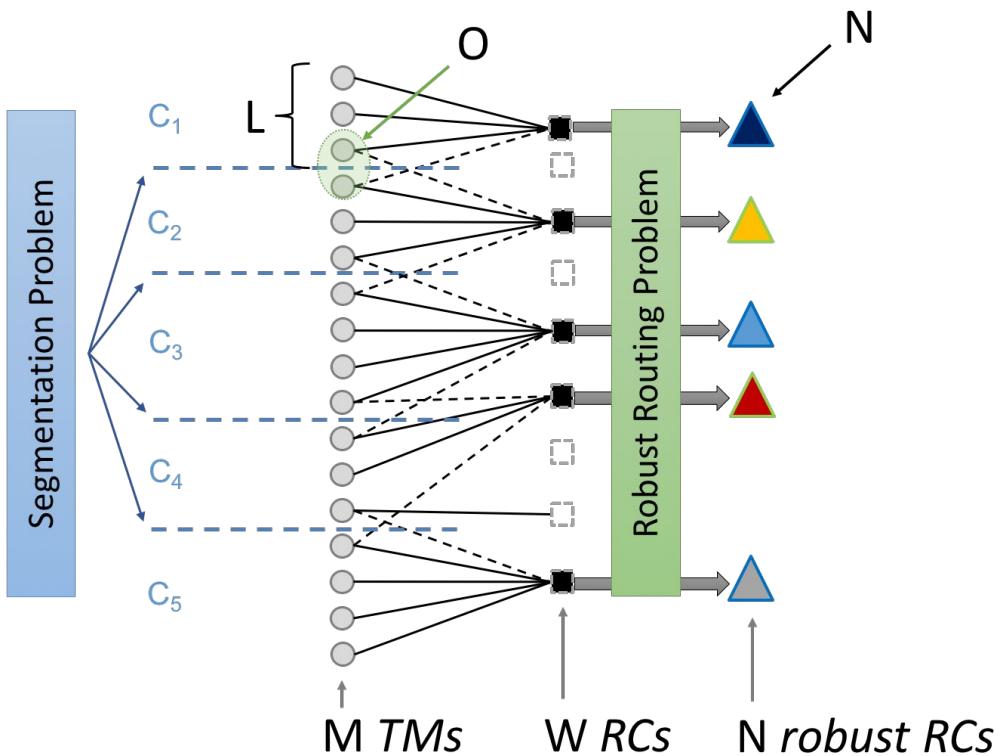
- Clustered Robust Routing (CRR)
  - Algorithm to build a set of robust routing (RR) configuration associated to clusters of TMs
  - Clustering of TM space in time, traffic and routing domains
  - Stability of configurations (guaranteed routing holding time)
- SDN controller logic
  - TM collection (e.g. for a 24h period)
  - CRR execution
  - Activation of RR config for the following 24h



# Clustered Robust Routing

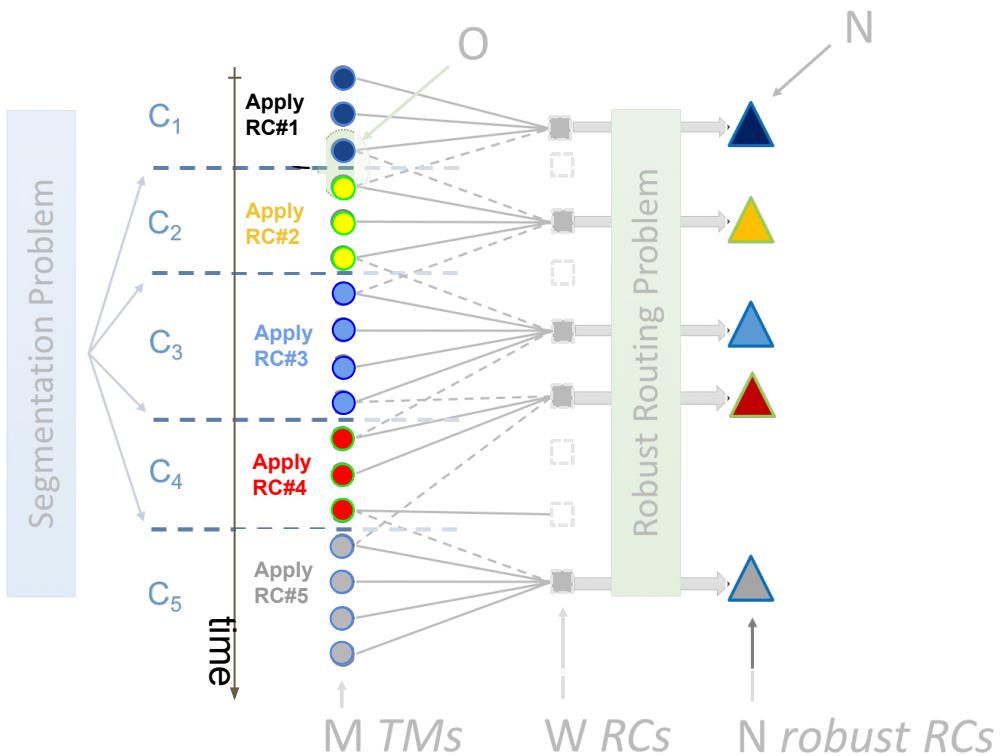
- Find the best assignment of **M** TMs to **N** robust RCs to find **N** TM clusters having minimum length of **L** TMs and an overlap of **O** TMs
- Joint optimization of routing and clusters
  - output set of RCs is required as input by the clustering logic
- Two-steps iterative algorithm
  - Segmentation Problem (**SP**)
  - Robust Routing Problem (**RRP**)

# CRR = SP + RRP



- INPUT
  - TM history/prediction
  - Set of  $W$  RCs
- PARAMETERS
  - $N$  = # of RCs
  - $L$  = cluster size
  - $O$  = cluster overlap amount
- OBJECTIVE FUNCTION:
  - Minimizing the sum of MLU of each TM over its assigned configuration
- OUTPUT
  - Set of  $N$  robust RCs
  - RCs activation times

# CRR = SP + RRP



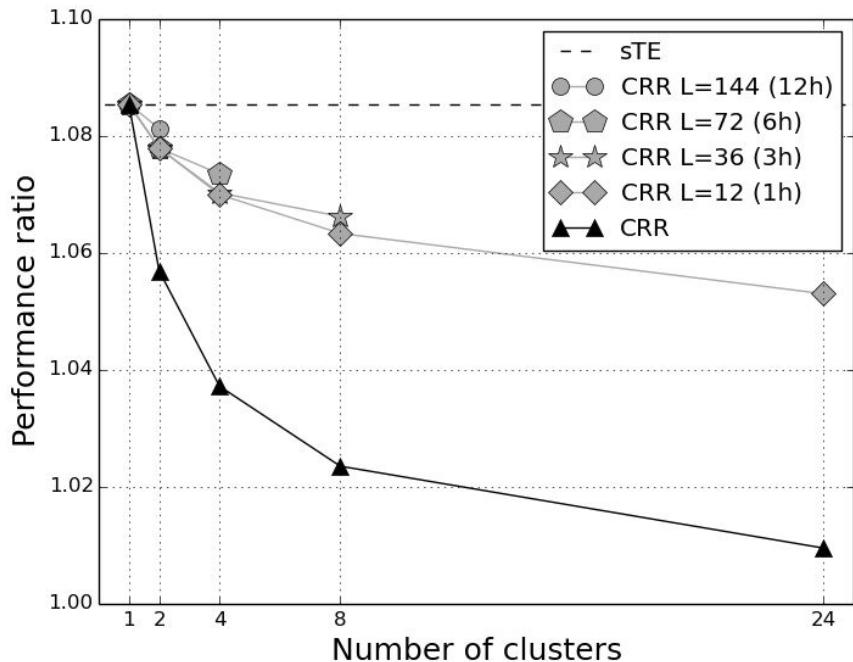
- **INPUT**
  - TM history/prediction
  - Set of  $W$  RCs
- **PARAMETERS**
  - $N$  = # of RCs
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- **OBJECTIVE FUNCTION:**
  - Minimizing the sum of MLU of each TM over its assigned configuration
- **OUTPUT**
  - Set of  $N$  robust RCs
  - RCs activation times

# Numerical evaluation

- Abilene backbone network
- 11 nodes
- 5-min granularity TMs
- CRR objective:  
minimize TM-averaged MLU
- Performance ratio with respect  
to Dynamic TE
- Results are averaged over 7 days

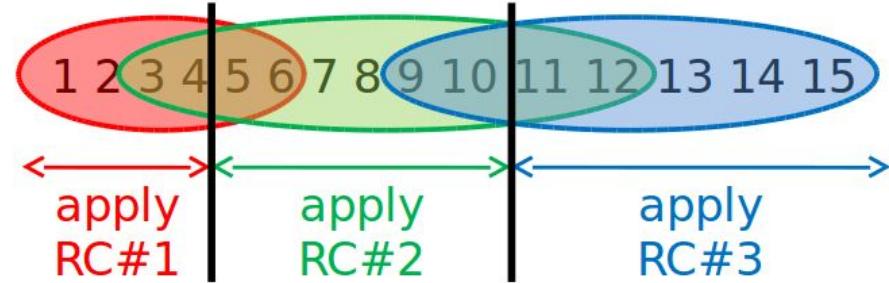
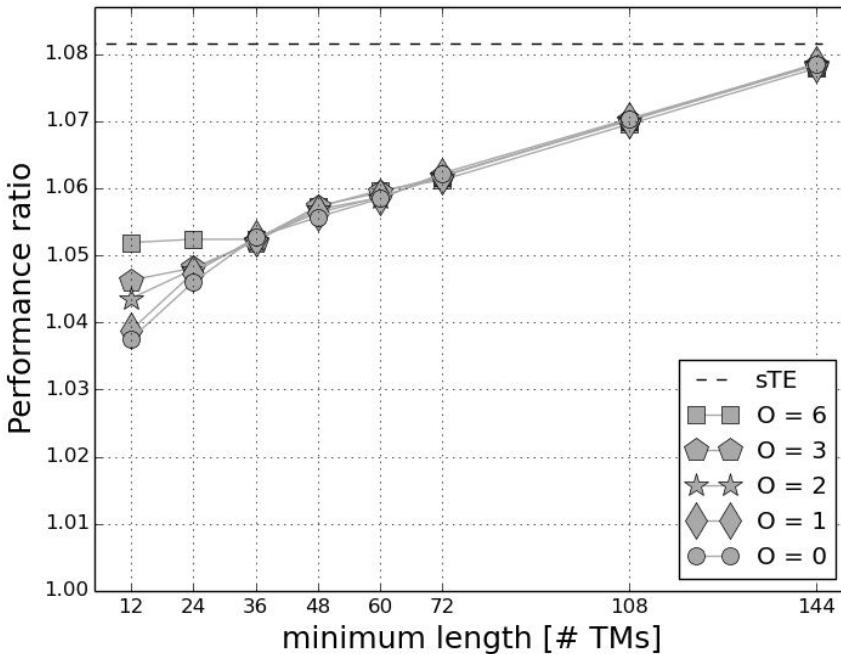


# Minimum cluster length



- Parameter **L** defines the minimum cluster size
  - translates in a guaranteed routing configuration duration
- It allows to tune the tradeoff between Stable TE and Dynamic TE

# Overlapping clusters



- Network reconfiguration is not instantaneous
  - Consistent updates mechanisms
- Clusters can be overlapped
  - Routing configurations take into account **O** TMs before/after the boundaries of the clusters
- Overlaps help slow consistent updates mechanism
- Negligible impact as **L** >> **O**

# Impact of TM prediction error

- We run the CRR over a noisy version of the TMs and applied its output to the original set of TMs
- Table reports % increase of objective function w.r.t. Dynamic TE
  - different levels of noise ( $\alpha$ )
  - different levels of robustness (L)
- Performance decrease
  - But limited to 10-12% wrt DynTE
- Larger clusters are better as noise increases
  - Resort to sTE if prediction quality is low

		$\alpha$	0	15	30	45	60
		cluster:	0	15	30	45	60
sTE		6.52	7.02	8.19	9.25	<b>10.52</b>	
CRR	L=72	4.07	5.13	6.93	<b>8.94</b>	11.09	
	L=60	4.04	5.23	7.24	9.44	11.19	
	L=48	3.76	5.06	7.12	9.61	11.79	
	L=36	3.17	4.55	<b>6.74</b>	9.04	11.33	
	L=24	2.84	4.50	6.85	9.35	11.93	
	L=12	<b>2.06</b>	<b>4.29</b>	7.04	10.08	12.28	

# Role of the SDN controller

- Offline
  - TM collection (e.g. for a 24h period)
  - CRR execution (e.g. during night)
- Online
  - estimation of current traffic scenario
  - activation of the proper RR configuration
  - handling of unexpected events
    - big traffic changes wrt planned scenarios
    - network failures

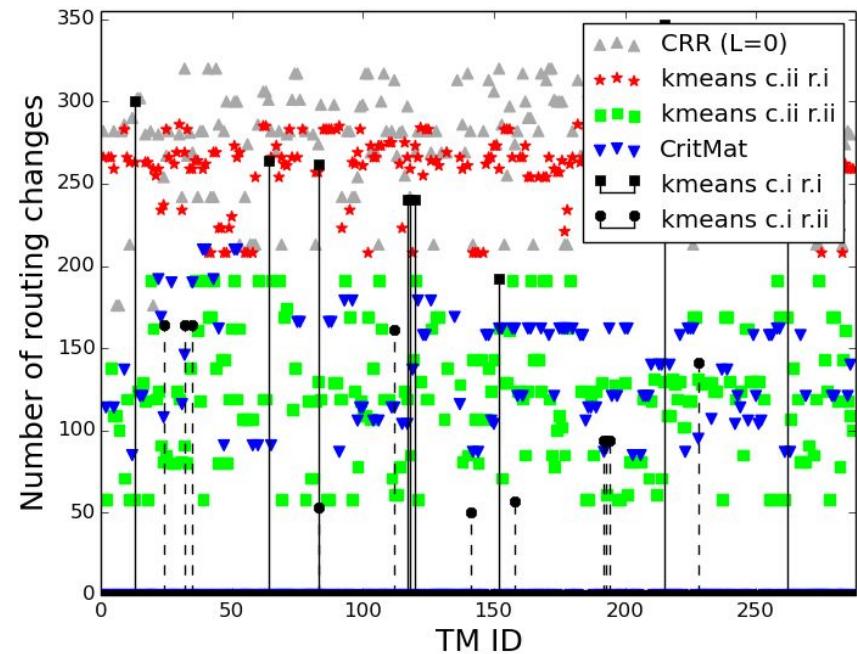
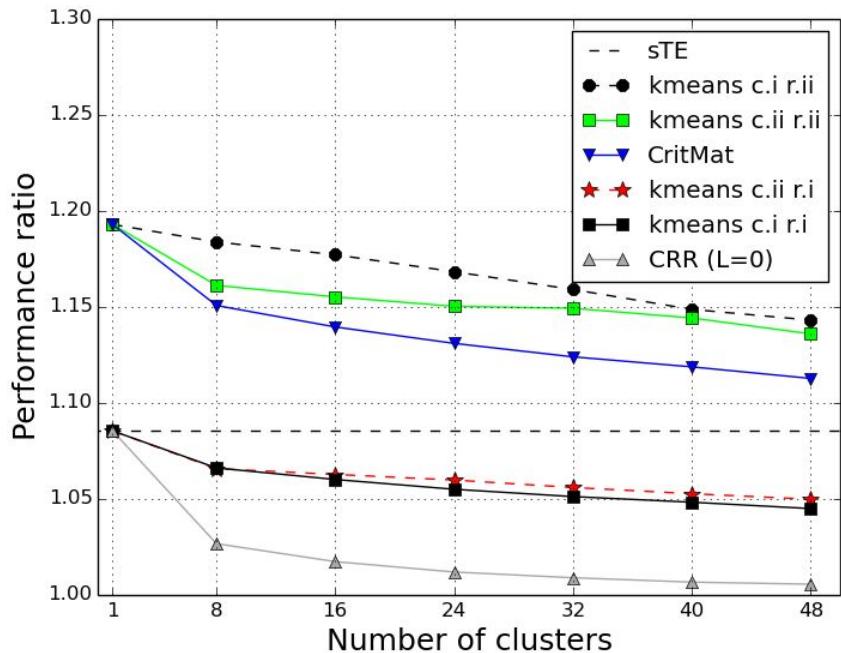
# Conclusion

- Clustered Robust Routing (CRR)
  - reduced number of routing configuration
  - guaranteed routing holding time
- SDN controller plays a key role
  - a-posteriori evaluation of TM prediction error
  - adaptive selection of the best level of robustness

# Thanks!

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# TM clustering and routing changes



SP

$$\begin{aligned} \min & \sum_{i \in \mathcal{T}, j \in \mathcal{R}} x_{ij} \delta_{i,j} + \\ & \frac{1}{2} \sum_{i \in \mathcal{T}, j \in \mathcal{R}} y_{ij} \left( \sum_{(i-O < k \leq i)_{|\mathcal{T}|}} \delta_{k,j} - \sum_{(i+1 \leq k \leq i+O)_{|\mathcal{T}|}} \delta_{k,j} \right) + \\ & \frac{1}{2} \sum_{i \in \mathcal{T}, j \in \mathcal{R}} w_{ij} \left( \sum_{(i \leq k < i+O)_{|\mathcal{T}|}} \delta_{k,j} - \sum_{(i-O \leq k < i)_{|\mathcal{T}|}} \delta_{k,j} \right) \end{aligned} \quad (1)$$

$$y_{ij} \geq x_{(i+1)|\mathcal{T}|j} - x_{ij}, \quad \forall i \in \mathcal{T}, j \in \mathcal{R} \quad (2)$$

$$\sum_{i \in \mathcal{T}} y_{ij} \leq z_j, \quad \forall j \in \mathcal{R} \quad (3)$$

$$\sum_{i \in \mathcal{T}, j \in \mathcal{R}} y_{ij} \leq \sum_{j \in \mathcal{R}} z_j \quad (4)$$

$$\sum_{j \in \mathcal{R}} x_{ij} = 1, \quad \forall i \in \mathcal{T} \quad (5)$$

$$\sum_{i \in \mathcal{T}} x_{ij} \geq L \cdot z_j, \quad \forall j \in \mathcal{R} \quad (6)$$

$$\sum_{j \in \mathcal{R}} z_j \leq N \quad (7)$$

$$x_{ij}, y_{ij}, z_j \in \{0, 1\}, \quad \forall i \in \mathcal{T}, j \in \mathcal{R} \quad (8)$$

$x_{ij}$	$j \in \mathcal{R}$		
	1	1	1
$i \in T$			
		1	1
			1
			1
			1
	1		

(a) Segmentation  
with no overlap

$x_{ij}$	$j \in \mathcal{R}$		
$i \in \mathcal{T}$	1		
	1		
		1	
		1	
		1	
			1
			1
			1
	1		

### (b) Segmentation with $O = 1$

$$[\mathbf{RR}] : \min. \gamma_{max} \quad \text{s. t.:} \quad (13)$$

$$\sum_{(i,j) \in \mathcal{L}} f_{ij}^h - \sum_{(j,i) \in \mathcal{L}} f_{ji}^h = \begin{cases} 1 & \text{if } i = O_h \\ -1 & \text{if } i = D_h \\ 0 & \text{otherwise} \end{cases}$$

$$\forall i \in \mathcal{N}, h \in \mathcal{H} \quad (14)$$

$$\sum_{h \in \mathcal{H}} d_h^m \cdot f_{ij}^h \leq c_{ij}, \forall m \in \mathcal{T}_c, (i, j) \in \mathcal{L} \quad (15)$$

$$\gamma_{max} \geq \frac{\sum_{h \in \mathcal{H}} d_h^m f_{ij}^h}{c_{ij}}, \forall m \in \mathcal{T}_c, (i, j) \in \mathcal{L} \quad (16)$$

$$0 \leq f_{ij}^h \leq 1, \quad \forall h \in \mathcal{H}, (i, j) \in \mathcal{L} \quad (17)$$