

Traffic Engineering

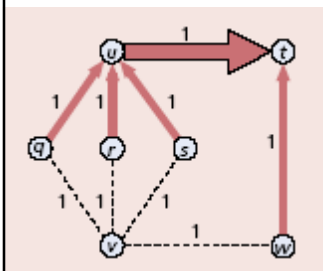
- ❑ configuring routes to traffic demands to
 - improve user performance
 - use network resources more efficiently
- ❑ operates at coarse timescales
 - not for failures, sudden traffic changes
- ❑ uses shortest path computations
 - OSPF, ISIS

Q: how to set link weights?

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Effect of link weights

- ❑ unit link weights
- ❑ local change to congested link
- ❑ global optimization
 - to balance link utilizations



Traffic Engineering Framework

- knowledge of topology
- traffic matrix
 - K - set of origin destination flows
 - $k \in K$, d_k - demand, s_k - source, t_k - destination
- how to get traffic matrix?
 - SNMP
 - edge measurements + routing tables
 - network tomography
 - packet sampling
- optimization criteria
 - minimize maximum utilization
 - keep utilizations below 60%

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How does one set link weights?

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Linear programming problem

Primal

$$\begin{aligned} &\text{minimize} && c^T x \\ &\text{subject to} && Ax = b \\ &&& x \geq 0, \end{aligned}$$

Dual

$$\begin{aligned} &\text{maximize} && y^T b \\ &\text{subject to} && y^T A \leq c^T \end{aligned}$$

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Complementary slackness

Let x and y be feasible solutions. A necessary and sufficient condition for them to be optimal is that for all i

1. $x_i > 0 \Rightarrow y^T A_i = c_i$
2. $x_i = 0 \Leftarrow y^T A_i < c_i$

Here A_i is i -th column of A

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Example: primal (P-SP)

- topology $G = (V, E)$, link weights $\{w_{ij} : (i, j) \in E\}$
- K - set of origin destination flows
 - $k \in K$, d_k - demand, s_k - source, t_k - destination
- X_{ij}^k fraction of flow k going over $(i, j) \in E$
- for $k \in K$

$$\begin{aligned} \min \quad & \sum_{(i,j) \in E} w_{ij} X_{ij}^k \\ \text{s.t.} \quad & \sum_{j:(i,j) \in E} X_{ij}^k - \sum_{j:(j,i) \in E} X_{ji}^k = 0, \quad i \neq s_k, t_k \\ & \sum_{j:(s_k,j) \in E} X_{ij}^k - \sum_{j:(j,s_k) \in E} X_{ji}^k = 1, \\ & \sum_{j:(t_k,j) \in E} X_{ij}^k - \sum_{j:(j,t_k) \in E} X_{ji}^k = -1, \\ & X_{ij}^k \geq 0 \end{aligned}$$

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Interpretation

- let $\{\bar{X}_{ij}^k\}$ be optimal solutions
- if $\{\bar{X}_{ij}^k\}$ takes values 0 and 1, corresponds to shortest paths
- if $\{\bar{X}_{ij}^k\}$ takes other values, there exist multiple shortest paths.

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Example: dual (D-SP)

$$\begin{aligned} \max \quad & \sum_{k \in K} U_{t_k}^k \\ \text{s.t.} \quad & U_j^k - U_i^k \leq w_{ij}, \quad k \in K, (i, j) \in E \\ & U_{s_k}^k = 0, \quad k \in K \end{aligned}$$

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Example

$$\begin{aligned} \max \quad & \sum_{k \in K} U_{t_k}^k \\ \text{s.t.} \quad & U_j^k - U_i^k \leq w_{ij}, \quad k \in K, (i, j) \in E \\ & U_{s_k}^k = 0, \quad k \in K \end{aligned}$$

- $\{\bar{U}_i^k\}$ optimal solution to dual problem
- $\bar{X}_{ij}^k > 0 \Rightarrow \bar{U}_j^k - \bar{U}_i^k = w_{ij}$, \bar{U}_j^k length of shortest path from s_k to j
- $\bar{U}_{t_k}^k$ length of shortest path from s_k to t_k

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Traffic engineering problem: minimize maximum link utilization

- topology $G = (V, E)$
- c_{ij} - capacity of link $(i, j) \in E$
- K - set of origin destination flows
 - $k \in K$, d_k - demand, s_k - source, t_k - destination
- α - maximum link utilization

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LP formulation

min α

$$s.t. \sum_{j:(i,j) \in E} X_{ij}^k - \sum_{j:(j,i) \in E} X_{ji}^k = \begin{cases} 0, & i \neq s_k, t_k, k \in K \\ 1, & i = s_k, k \in K \\ -1, & i = t_k, k \in K \end{cases}$$

$$\sum_{k \in K} d_k X_{ij}^k \leq c_{ij} \alpha, \quad (i, j) \in E$$

$$X_{ij}^k \geq 0$$

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LP formulation

$$\min \alpha + r \sum_{k \in K} \sum_{(i,j) \in E} X_{ij}^k$$

$$s.t. \sum_{j:(i,j) \in E} X_{ij}^k - \sum_{j:(j,i) \in E} X_{ji}^k = \begin{cases} 0, & i \neq s_k, t_k, k \in K \\ 1, & i = s_k, k \in K \\ -1, & i = t_k, k \in K \end{cases}$$

$$\sum_{k \in K} d_k X_{ij}^k \leq c_{ij} \alpha, \quad (i,j) \in E$$

$$X_{ij}^k \geq 0$$

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LP formulation

- can be many solutions with same α
- in case of tie, want solution with short paths

⇒ add term $r \sum_{k \in K} \sum_{(i,j) \in E} X_{ij}^k$
with small r to cost

- use standard LP algorithms (Simplex) to solve

Q: can we find link weights so that solution comes from shortest path problem?

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Duality revisited

Primal

$$\begin{aligned} &\text{minimize} && c^T x \\ &\text{subject to} && Ax = b_1 \\ &&& A'x \geq b_2 \\ &&& x \geq 0, \end{aligned}$$

- free variables in primal \Rightarrow equality constraints in dual

Dual

$$\begin{aligned} &\text{maximize} && y_1^T b_1 + y_2^T b_2 \\ &\text{subject to} && y_1^T A + y_2^T A' \leq c^T \\ &&& y_2 \geq 0 \end{aligned}$$

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Dual formulation

- decision variables $\{U_i^k\}, \{W_{ij}\}$

$$\begin{aligned} &\max \sum_{k \in K} d_k U_{t_k}^k \\ &s.t. U_j^k - U_i^k \leq W_{ij} + r, \quad k \in K, (i, j) \in E \\ &\sum_{(i, j) \in E} c_{ij} W_{ij} = 1, \\ &W_{ij} \geq 0, U_{s_k}^k = 0 \end{aligned}$$

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Properties of primal-dual solutions

- optimal solution to primal problem $\{\bar{X}_{ij}^k\}, \bar{\alpha}$
dual problem $\{\bar{U}_i^k\}, \{\bar{W}_{ij}\}$
- if $\bar{X}_{ij}^k > 0$, then $\bar{U}_j^k - \bar{U}_i^k = \bar{W}_{ij} + r$
- can think of \bar{U}_j^k as shortest path distance from s_k to j when link weights are $\{\bar{W}_{ij} + r\}$

Therefore: solution to TE problem is also solution to shortest path problem with

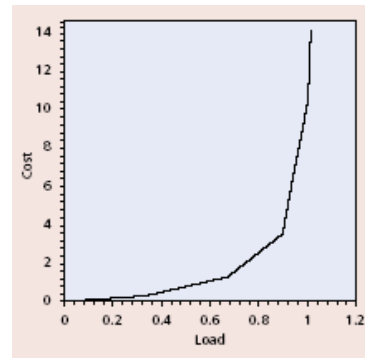
$$w_{ij} = \bar{W}_{ij} + r$$

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Link weight assignment

- works for rich set of cost functions
- example:

$$\Phi = \sum_{(i,j) \in E} \Phi_{ij} \left(\sum_{k \in K} d_k X_{ij}^k \right)$$
- where Φ_{ij} are piecewise linear



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Issues

- ❑ solutions are flow specific - need destination specific solutions
 - not a big deal, can reformulate to account for this
- ❑ solutions may not support equal split rule of OSPF
 - accounting for this yields NP-hard problem
 - see heuristics in FT paper
 - modify IP routing

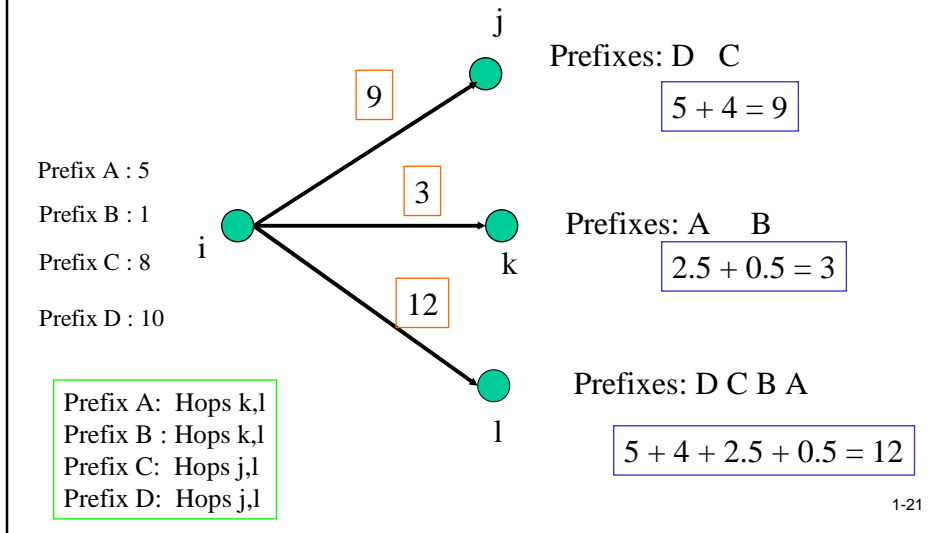
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One approach to overcome the "splitting problem"

- ❑ current routing tables have thousands of routing prefixes
- ❑ instead of routing each prefix on all equal cost paths, selectively assign next hops to (each) prefix
 - i.e., remove some equal cost next hops assigned to prefixes
- ❑ goal: to approximate optimal link load

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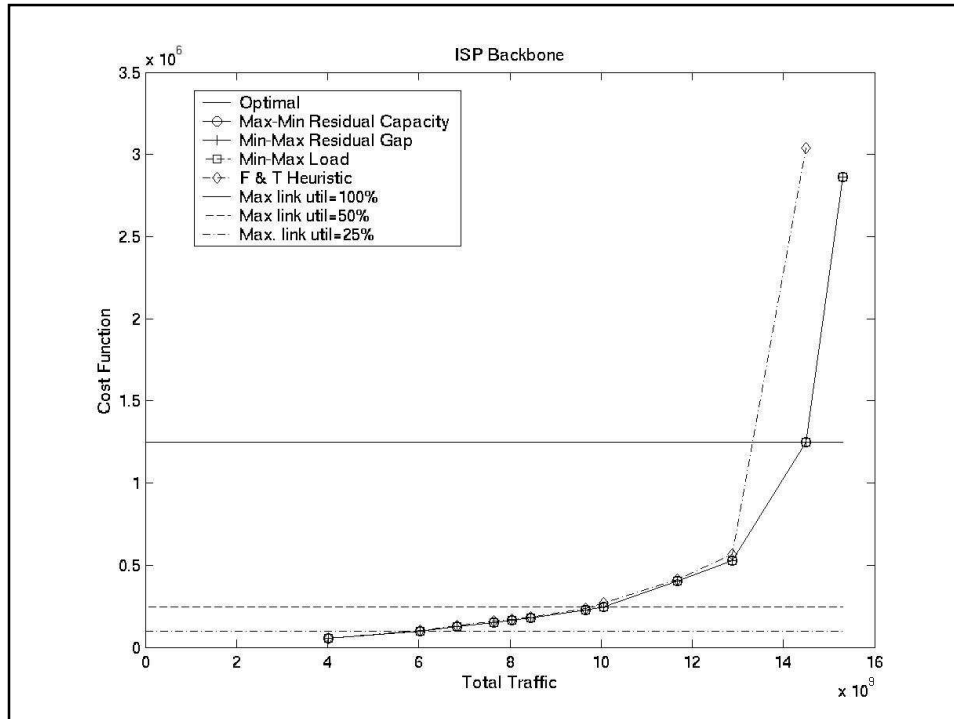
Example : EQUAL-SUBSET-SPLIT



Advantages

- ❑ requires no change in data path
- ❑ can leverage existing routing protocols
- ❑ current routers have 10,000s of routes in routing tables
 - provides large degree of flexibility in next hop allocation to match optimal allocation

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Summary

- can use OSPF/ISIS to support traffic engineering objectives
- performance objectives link weights
- equal splitting rule complicates problem
 - heuristics provide good performance
 - small changes to IP routing provide in better performance
- MPLS suffers none of these problems

Recent progress on MPLS TE

□ Oblivious routing

- Goal: find a routing R that minimizes the performance ratio under worst traffic matrix
 - $PR = \text{MLU under R} / \text{MLU under optimal routing}$
- Result: on realistic ISP topologies, can achieve PR of ~ 2

□ COPE [sigcomm06]

- Common-case Optimization with Penalty Envelope
- Achieve close to optimal MLU with normal traffic
- Provide worst-case PR bound with unexpected traffic

□ R3: Resilient Routing Reconfiguration [ongoing]

- Goal: avoid congestion under fast rerouting
- Idea: convert topology uncertainty into traffic uncertainty

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